

# Research Reports on Mathematical and Computing Sciences

Level-wise Node Size Distribution  
of Randomly Generated Regular Trees

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March 2010, C-269

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SERIES C: **Computer Science**

# 1 Introduction

For analyzing a random graph model for explaining a hierarchical clique structure of large scale Web networks, some statistical properties of random regular trees have been used in [SUW10]. In this note, we give a detailed analysis of these properties.

We consider random  $k$ -regular trees for any integer  $k \geq 2$  that will be fixed throughout this paper. We consider a branching process that has been known as *Galton-Watson process*. For a given parameter  $\mu_0$ ,  $0 < \mu_0 < 1$ , the process, starting from an initial node, generates a tree in the following way:

1. For each *open node*  $v$ ,
  - (a) with probability  $p_k = \mu_0/k$ , create  $k$  new open nodes, add them to  $v$  as its child nodes, and change the status of  $v$  to *closed*,
  - (b) otherwise (i.e., with probability  $1 - p_k$ ), change the status of  $v$  to *closed* without adding any child nodes.
2. Repeat the above until all nodes are closed.

Let  $T$  denote a tree generated by this process. The initial node is called a *root* node and a node with no children is called a *leaf* node. For each node  $v$  of  $T$ , we define its height  $h(v)$  and level  $l(v)$  inductively as follows.

$$h(v) = \begin{cases} 0, & \text{if } v \text{ is a root node, and} \\ h(v') + 1, & \text{otherwise, and where } v' \text{ is the parent node of } v; \end{cases}$$
$$l(v) = \begin{cases} 0, & \text{if } v \text{ is a leaf node, and} \\ \max\{l(v_1), \dots, l(v_k)\} + 1, & \text{otherwise, and where } v_1, \dots, v_k \text{ are child nodes of } v. \end{cases}$$

The height of a tree is the maximum height of nodes in  $T$ . Note that the height of a tree equals the level of the root node of the tree.

The height of  $T$  as well as the number of nodes with a given height  $h$  have been studied in depth in the literature (see, e.g., [Agr74]). On the other hand, less is known about the number of nodes of given level  $l$ . The purpose of this note is to show reasonable upper and lower bounds for the expected number of nodes of given level  $l$ .

# 2 Analysis

Let  $T$  denote a random  $k$ -regular tree generated by the above process. In the following, we assume that  $T$  is finite; thus, precisely speaking, probabilities and expectations discussed below are all conditional on the fact that  $T$  is finite. Recall that we assume that  $kp_k = k(\mu_0/k) = \mu_0 < 1$ ; on the other hand, it has been known (see, e.g., [Fel68]) that  $T$  is finite with probability 1 in this case.

Fix any  $l \geq 0$ . Let  $M(l)$  denote the expected number of nodes with level  $l$  in  $T$ . Our goal is to give good upper and lower bounds for  $M(l)$ . For this, we use  $P(l)$ , the probability that the root has level  $l$ , i.e. the depth of  $T$  is  $l$ .

We analyze  $M(l)$  by estimating all possible contributions to it. First, consider the case that the root has level  $l$ . If the root has level  $l$ , other nodes cannot have level  $l$ , so there is only one level  $l$  node in  $T$ . The root has level  $l$  with probability  $P(l)$ ; hence, this contributes  $P(l) \cdot 1$  to  $M(l)$ . Then consider the other case. Since  $M(l)$  would be 0 for  $l \geq 1$  if the root were not expanded, consider the situation that the root is expanded (which occurs with probability  $p_k$ ). Let  $v_1, \dots, v_k$  denote the child nodes of the root and let  $T_1, \dots, T_k$  denote the trees rooted by these nodes. Then the contributions from  $T_1, \dots, T_k$  are  $p_k$  times the expected number of level  $l$  nodes of those trees. Each  $T_i$  follows the same probability distribution as  $T$ ; thus, we may use  $M(l)$  for the expected number of level  $l$  nodes of  $T_i$ . Since these are all contributions, we have

$$M(l) = P(l) + p_k \cdot k \cdot M(l),$$

and, since we assumed that the number of nodes on the tree  $T$  is finite, this implies that

$$M(l) = \frac{P(l)}{1 - p_k k} = \frac{P(l)}{1 - \mu_0}. \quad (1)$$

Now our task is to estimate  $P(l)$ , and we will discuss it in the rest of this note. Let  $g(z)$  denote the probability generating function (p.g.f.) of the number of children of a node in our process; that is,  $g(z) = 1 - p_k + p_k z^k$ . Note that  $g'(1) = \mu_0$  is the expected number of children of one node and that we assumed  $\mu_0 < 1$ . The p.g.f. of the number of nodes with height  $i$ , denoted by  $Z_i$ , is  $g_i(z)$  where  $g_1(z) = g(z)$  and  $g_j(z) = g(g_{j-1}(z))$  for  $j > 1$  [Fel68]. However, it is hard to obtain the closed-form of  $g_i(z)$ .

Let  $q(l)$  denote the probability that the root has level at least  $l$ . We here note some basic equations of  $P(l)$  and  $q(l)$ .

$$P(l) = q(l) - q(l+1) \quad (\text{for } l \geq 0) \quad (2)$$

$$q(l) = p_k \left\{ 1 - (1 - q(l-1))^k \right\} \quad (\text{for } l \geq 1) \quad (3)$$

$$q(l) < \mu_0 q(l-1). \quad (\text{for } l \geq 1) \quad (4)$$

Bound (4) is derived from (3) as follows:

$$q(l) = p_k \left\{ 1 - (1 - q(l-1))^k \right\} < p_k \left\{ 1 - (1 - kq(l-1)) \right\} = \mu_0 q(l-1).$$

For an upper bound of  $P(l)$ , we have the following Lemma.

**Lemma 1.** We have  $P(0) = 1 - p_k$  and  $P(1) = p_k(1 - p_k)^k = \frac{\mu_0}{k} \left(1 - \frac{\mu_0}{k}\right)^k$ . For any  $l > 1$ , we have

$$P(l) < \frac{\mu_0^l}{k} \left(1 - \frac{\mu_0}{k}\right)^k.$$

**Proof.** By definition,  $P(0)$  and  $P(1)$  are the probability that the root node has level 0 and 1 respectively, so we immediately have  $P(0) = 1 - p_k$  and  $P(1) = p_k(1 - p_k)^k$ . For any  $0 < x < y < 1$ , it is easy to show that

$$(1 - x)^k - (1 - kx) < (1 - y)^k - (1 - ky).$$

Using this with (2) and (3), we have

$$\begin{aligned} P(l) &= q(l) - q(l+1) = p_k \left[ \left\{ 1 - (1 - q(l-1))^k \right\} - \left\{ 1 - (1 - q(l))^k \right\} \right] \\ &< p_k \left[ \left\{ 1 - (1 - kq(l-1)) \right\} - \left\{ 1 - (1 - kq(l)) \right\} \right] \\ &= p_k k (q(l-1) - q(l)) = \mu_0 P(l-1). \end{aligned}$$

Hence we obtained  $P(l) < \mu_0^{l-1} P(1) = \frac{\mu_0^l}{k} (1 - \frac{\mu_0}{k})^k$ . □

For analyzing a lower bound of  $P(l)$ , we need both upper and lower bounds of  $q(l)$ . An upper bound is derived inductively from (4). Noting that  $q(1) = p_k = \frac{\mu_0}{k}$ , we have

$$q(l) < \frac{\mu_0^l}{k}. \tag{5}$$

For showing a lower bound of  $q(l)$ , we make use of facts that have been shown in the literature. Note first that  $q(l)$  satisfies the following relationships with the p.g.f.  $g_l(z)$ :

$$\begin{aligned} 1 - q(l) &= \Pr[\text{the level of the root node} < l] \\ &= \Pr[\text{the number of nodes with height } l \text{ is } 0] \\ &= \Pr[Z_l = 0] = g_l(0). \end{aligned}$$

However, as mentioned before, the closed-form of  $g_l(z)$  is hard to obtain. In [Agr74], Agresti used a fractional linear generating function (f.l.g.f.) to obtain good upper/lower bounds of  $g_l(z)$ . We follow their analysis and obtain the following lower bound.

**Lemma 2.** For any  $l \geq 1$ , we have

$$q(l) > \frac{\mu_0^l}{k} (1 - \mu_0).$$

**Proof.** For any p.g.f.  $g(z)$ , let  $U(z)$  be any p.g.f. satisfying  $g(z) \leq U(z)$  for  $0 \leq z \leq 1$ . We first recall the following fact shown by Seneta (Lemma A of [Sen67]).

**Fact 1.** For any  $l \geq 1$ , and for any  $0 \leq z \leq 1$ , we have

$$g_l(z) \leq U_l(z),$$

where  $U_l$  is defined inductively by  $U_l(z) = U(U_{l-1}(z))$  and  $U_1(z) = U(z)$ .

**Proof.** Since  $U_l(z)$  is a p.g.f., it is an increasing function; also since  $g(z)$  is a p.g.f., it satisfies  $0 \leq g(z) \leq 1$  for any  $0 \leq z \leq 1$ . Thus we have

$$\begin{aligned} g_l(z) &= g_{l-1}(g(z)) \\ &\leq U_{l-1}(g(z)) \quad (\text{by induction}) \\ &\leq U_{l-1}(U(z)) \quad (U_{l-1}(z) \text{ is increasing}) \\ &= U_l(z). \end{aligned}$$

□ (Fact 1)

Thus, by using some appropriate  $U(z)$ , we can give the following lower bound of  $q(l)$ :

$$\begin{aligned} 1 - q(l) &= \Pr[\text{the level of the root} < l] \\ &= \Pr[\text{the number of nodes with height } l \text{ is } 0] \\ &= g_l(0) \leq U_l(0). \end{aligned} \tag{6}$$

For  $U(z)$ , we use the following fractional linear generating function (f.l.g.f.) introduced by Agresti ([Agr74], Lemma 3 (i)).

**Fact 2.** Define  $U(z)$  by

$$U(z) = 1 - p_k + \frac{p_k z}{k - (k-1)z}.$$

Then,  $U(z)$  satisfies  $g(z) \leq U(z)$  for any  $0 \leq z \leq 1$ .

**Proof.** By definition of  $g(z)$  and  $U(z)$ , it suffices to show

$$g(z) = 1 - p_k + p_k z^k \leq 1 - p_k + \frac{p_k z}{k - (k-1)z}$$

for all  $z$ ,  $0 \leq z \leq 1$ . This holds if and only if

$$t(z) = 1 - kz^{k-1} + (k-1)z^k = 1 - z^k - k(1-z)z^{k-1} \geq 0$$

for all  $z$ ,  $0 \leq z \leq 1$ . Note that  $t(1) = 0$  and that

$$\begin{aligned} t'(z) &= -kz^{k-1} - k(k-1)z^{k-2}(1-z) + kz^{k-1} \\ &= -k(k-1)z^{k-2}(1-z) \leq 0 \end{aligned}$$

for all  $z$ ,  $0 \leq z \leq 1$ . Hence,  $t(z) \geq 0$  for  $0 \leq z \leq 1$ .

□ (Fact 2)

Since  $U(z)$  is a f.l.g.f., we can obtain the closed form of  $U_l(z)$ , the  $l$ th iterate of  $U(z)$ , which is stated as follows (see Appendix for its derivation):

$$U_l(z) = 1 + \frac{\mu_0^l (1 - \mu_0)(z - 1)}{(k-1)(\mu_0^l - 1)z + (k - \mu_0 - (k-1)\mu_0^l)}.$$

Thus from (6) it follows

$$\begin{aligned} 1 - q(l) &\leq U_l(0) = 1 - \frac{\mu_0^l(1 - \mu_0)}{k - \mu_0 - (k - 1)\mu_0^l} \\ &< 1 - \frac{\mu_0^l(1 - \mu_0)}{k}, \end{aligned}$$

and hence

$$q(l) > \frac{\mu_0^l}{k}(1 - \mu_0).$$

□ (Lemma 2)

By (5) and Lemma 2,  $q(l)$  can be represented as

$$q(l) = \frac{\mu_0^l}{k}(1 - \mu_0) + \epsilon_l,$$

where  $0 < \epsilon_l < \frac{\mu_0^{l+1}}{k}$ . Now by (4), we have

$$\frac{\mu_0^{l+1}}{k}(1 - \mu_0) + \epsilon_{l+1} = q(l + 1) < \mu_0 q(l) = \mu_0 \left( \frac{\mu_0^l}{k}(1 - \mu_0) + \epsilon_l \right).$$

Hence we have  $\epsilon_{l+1} < \mu_0 \epsilon_l < \epsilon_l$ , from which it follows  $\epsilon_l - \epsilon_{l+1} > 0$ . Thus, we have

$$\epsilon_l - \epsilon_{l+1} = q(l) - q(l + 1) - \left\{ \frac{\mu_0^l}{k}(1 - \mu_0) - \frac{\mu_0^{l+1}}{k}(1 - \mu_0) \right\} > 0.$$

From this bound, we obtain the following lower bound of  $P(l)$ :

$$P(l) = q(l) - q(l + 1) > \left\{ \frac{\mu_0^l}{k}(1 - \mu_0) - \frac{\mu_0^{l+1}}{k}(1 - \mu_0) \right\} = \frac{\mu_0^l}{k}(1 - \mu_0)^2.$$

Then from this bound and Lemma 1, we obtain the following upper and lower bound of  $P(l)$ :

$$\frac{\mu_0^l}{k}(1 - \mu_0)^2 < P(l) < \frac{\mu_0^l}{k} \left( 1 - \frac{\mu_0}{k} \right)^k.$$

We now obtained both upper and lower bounds of  $P(l)$ , using them we have

$$\frac{\mu_0^l}{k}(1 - \mu_0) < M(l) < \frac{\mu_0^l}{k} \left( 1 - \frac{\mu_0}{k} \right)^k \frac{1}{1 - \mu_0} < \frac{\mu_0^l}{k} \frac{1}{1 - \mu_0}.$$

From this, the following Theorem is derived.

**Theorem 3.** Let  $C_1 = 1 - \mu_0$  and  $C_2 = \frac{1}{1 - \mu_0}$ . Then for any  $l \geq 0$ , we have

$$C_1 \mu_0^l \frac{1}{k} < M(l) < C_2 \mu_0^l \frac{1}{k}.$$

### 3 Concluding remarks

In this note, we discuss a branching process and give detail analysis for the expected number of nodes with level  $l$ . We focus on the special p.g.f.  $g(z) = 1 - p_k + p_k z^k$ . Many detailed analysis of  $P(l)$  and  $q(l)$  of other p.g.f. were given in the literature, e.g., [Fel68, AN72, Har63], so we can apply these analysis to Equation 1, and obtain the expected number of nodes with level  $l$ .

### References

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### Appendix

We here derive the  $l$ th iteration of  $U(z)$ . Let us recall our definition of  $U(z)$ , that is,

$$U(z) = 1 - p_k + \frac{p_k z}{k - (k-1)z} = \frac{(k-1-\mu_0)z - (k-\mu_0)}{(k-1)z - k}.$$

Also recall that its  $l$ th iteration  $U_l(z)$  is defined inductively by  $U_l(z) = U(U_{l-1}(z))$  for  $l > 1$  and  $U_1(z) = U(z)$ .

To derive  $U_l(z)$ , we use a linear function  $L(z) = az+b$  and  $f(z) = L^{-1}(U(L(z)))$ . Due to the following lemma, for evaluating  $U_l(z)$ , it suffices to get good  $a$  and  $b$  such that  $f_l(z)$  is easily calculated.

**Lemma 4.**

$$U_l(z) = L(f_l(L^{-1}(z))).$$

**Proof.** By  $f(z) = L^{-1}(U(L(z)))$ , we have  $U(z) = L(f(L^{-1}(z)))$ . Then we prove the lemma by induction. We already have it for  $l = 1$ . Let us assume that  $U_l(z) = L(f_l(L^{-1}(z)))$ . Then we have

$$\begin{aligned} U_{l+1}(z) &= U(U_l(z)) = L(f(L^{-1}(U_l(z)))) \\ &= L(f(L^{-1}(L(f_l(L^{-1}(z)))))) = L(f(f_l(L^{-1}(z)))) \\ &= L(f_{l+1}(L^{-1}(z))). \end{aligned}$$

□

Let  $a = \frac{1-\mu_0}{k-1}$  and  $b = 1$ ; then we have

$$\begin{aligned} f(z) &= L^{-1}(U(L(z))) = \frac{1}{a}(U(az+1)-1) \\ &= \frac{a(k-1-\mu_0)z + (k-1-\mu_0) - (k-\mu_0) - \{a(k-1)z + (k-1) - k\}}{a\{a(k-1)z + (k-1) - k\}} \\ &= \frac{a(k-1-\mu_0)z - 1 - a(k-1)z + 1}{a\{a(k-1)z - 1\}} \\ &= \frac{-\mu_0 z}{a(k-1)z - 1} = \frac{-\mu_0 z}{(1-\mu_0)z - 1} \quad (\text{by } a = \frac{1-\mu_0}{k-1}) \\ &= \frac{z}{\left(1 - \frac{1}{\mu_0}\right)z + \frac{1}{\mu_0}}. \end{aligned}$$

**Lemma 5.** Let  $K = \frac{1}{\mu_0}$ . Then we have

$$f_l(z) = \frac{z}{K^l + (1-K^l)z}.$$

**Proof.** For  $l = 1$ , we have

$$f_1(z) = \frac{z}{\frac{1}{\mu_0} + \left(1 - \frac{1}{\mu_0}\right)z} = \frac{z}{K + (1-K)z},$$

and the lemma holds. For  $l \geq 1$ , we prove by induction as follows:

$$\begin{aligned} f_{l+1}(z) &= \frac{f_l(z)}{K + (1-K)f_l(z)} = \frac{\frac{z}{K^l + (1-K^l)z}}{K + (1-K)\frac{z}{K^l + (1-K^l)z}} \\ &= \frac{z}{K^{l+1} + (1-K^l)Kz + (1-K)z} = \frac{z}{K^{l+1} + (1-K^{l+1})z} \end{aligned}$$

□

We now have the closed form of  $f_l(z)$ . That is,

$$f_l(z) = \frac{z}{K^l + (1-K^l)z} = \frac{z}{\left(\frac{1}{\mu_0}\right)^l + \left(1 - \left(\frac{1}{\mu_0}\right)^l\right)z} = \frac{\mu_0^l}{\left(\frac{1-z}{z}\right) + \mu_0^l}.$$



By using Lemma 4, we obtain the closed form of  $U_l(z)$  as follows:

$$\begin{aligned}
U_l(z) &= L(f_l(L^{-1}(z))) = a \left( f_l \left( \frac{z-1}{a} \right) \right) + 1 \\
&= a \frac{\mu_0^l}{\left( \frac{a+1-z}{z-1} \right) + \mu_0^l} + 1 \\
&= 1 + \frac{a\mu_0^l z - a\mu_0^l}{(\mu_0^l - 1)z + (a + 1 - \mu_0^l)} \\
&= 1 + \frac{\mu_0^l(1 - \mu_0)(z - 1)}{(k - 1)(\mu_0^l - 1)z + (k - \mu_0 - (k - 1)\mu_0^l)}.
\end{aligned}$$