Fitting the WHOIS Internet data

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This short technical manuscript contains supporting information for Ref. [1]. We consider the RIPE WHOIS internet data as characterized by the Cooperative Association for Internet Data Analysis (CAIDA) [2], and show that the Tempered Preferential Attachment (TPA) model [1] provides an excellent fit to this data. First we define the complementary cumulative probability distribution (ccdf), and then derive the ccdf for a TPA graph. Next we discuss the ccdf for the WHOIS data. Finally we discuss the fit provided by the TPA model and by a power law with exponential decay (PLED).

I. DEFINING THE CCDF

The complementary cumulative probability distribution, ccdf(x):

$$\mathrm{ccdf}(x) = 1 - \sum_{j=1}^{x-1} p_j = \sum_{j=x}^{\infty} p_j. \tag{1}$$

II. THE CCDF PREDICTED BY TPA WITH $A_1 \neq A_2$

A. First recall the recursion relations

The recursion relations defining the degree distribution for TPA graphs were derived explicitly in Refs. [3] and [4]. Here we derive the corresponding ccdf. These are Eqn's (16) and (17) in [3]:

$$p_i = \left(\prod_{k=2}^{i} \frac{k-1}{k+w} \right) p_1 = \left(\prod_{k=1}^{i-1} \frac{k}{k+w+1} \right) p_1, \quad \text{for} \quad i \leq A_2, \tag{2}$$

and

$$p_i = \left(\frac{A_2}{A_2+w} \right)^{i-A_2} p_{A_2} \;=\; q^{i-A_2}\, p_{A_2}, \quad \text{for} \quad i \geq A_2. \tag{3}$$

Note

$$p_{A_2} = \left(\prod_{k=1}^{A_2-1} \frac{k}{k+w+1} \right) p_1, \tag{4}$$

and, for convenience, we defined:

$$q \equiv \left(\frac{A_2}{A_2 + w}\right). \tag{5}$$

We will first calculate the CCDF for $i \geq A_2$ as we will use that result to determine the CCDF for $i < A_2$.

B. Calculating the CCDF, for $x \geq A_2$

Recall the definition of the CCDF from Eqn. (1):

$$\begin{aligned}
\mathrm{ccdf}(x) &= \sum_{j=x}^{\infty} p_j \\
&= p_{A_2} \sum_{j=x}^{\infty} q^{j-A_2} \\
&= p_{A_2} \sum_{j=0}^{\infty} q^{j+x-A_2} \\
&= p_{A_2} q^{x-A_2} \sum_{j=0}^{\infty} q^j.
\end{aligned} \tag{6}$$

Since $q < 1$, the sum in Eqn. (6) is a geometric series; $\sum_{j=0}^{\infty} q^j = 1/(1-q)$. Thus we can write:

$$\boxed{\mathrm{ccdf}(x) = \left(\frac{p_{A_2}}{1-q}\right) q^{x-A_2}, \quad \text{for} \quad x \geq A_2.} \tag{7}$$

C. Calculating the CCDF, for $x < A_2$

This is slightly more complicated, as we have different functional forms for $x < A_2$ and $x > A_2$.

$$\begin{aligned}
\mathrm{ccdf}(x) &= \sum_{j=x}^{\infty} p_j \\
&= \sum_{j=x}^{A_2-1} p_j + \sum_{j=A_2}^{\infty} p_j \\
&= \sum_{j=x}^{A_2-1} p_j + \mathrm{ccdf}(A_2) \\
&= \sum_{j=x}^{A_2-1} p_j + \left(\frac{p_{A_2}}{1-q}\right).
\end{aligned} \tag{8}$$

Plugging in the relation for p_i from Eqn. (3), we obtain:

$$\boxed{\mathrm{ccdf}(x) = p_{A_2}\left(\frac{1}{1-q} + \sum_{j=x}^{A_2-1}\prod_{k=j}^{A_2-1}\frac{k+w+1}{k}\right), \quad \text{for} \quad x < A_2.} \tag{9}$$

D. Standard Normalization

First we can check that Eqns. (7) and (9) give the same value for ccdf(A_2). They do:

$$\mathrm{ccdf}(A_2) = \frac{p_{A_2}}{1-q}. \tag{10}$$

And we can determine the value of p_{A_2} by the normalization condition that

$$\mathrm{ccdf}(1) = 1 = p_{A_2}\left(\frac{1}{1-q} + \sum_{j=1}^{A_2-1}\prod_{k=j}^{A_2-1}\frac{k+w+1}{k}\right). \tag{11}$$

In other words,

$$\boxed{p_{A_2} = \left(\frac{1}{1-q} + \sum_{j=1}^{A_2-1}\prod_{k=j}^{A_2-1}\frac{k+w+1}{k}\right)^{-1}.} \tag{12}$$

E. Normalizing without degree $d = 1$ nodes

We may want to neglect nodes with degree $d < 2$ for various reasons. In that case, the normalization would be:

$$\mathrm{ccdf}(2) = 1 = p_{A_2}\left(\frac{1}{1-q} + \sum_{j=2}^{A_2-1}\prod_{k=j}^{A_2-1}\frac{k+w+1}{k}\right). \tag{13}$$

Thus

$$\boxed{p_{A_2} = \left(\frac{1}{1-q} + \sum_{j=2}^{A_2-1}\prod_{k=j}^{A_2-1}\frac{k+w+1}{k}\right)^{-1}} \tag{14}$$

with Eqns. (7) and (9) unchanged (except Eqn. (9) now holds for $2 \le x < A_2$, rather than for $1 \le x < A_2$).

III. THE WHOIS CCDF, FOR $d > 1$

A. Whois data, renormalize to remove $d < 2$

By definition:

$$\sum_{j=1}^{\infty} p_j = 1.$$

Thus:

$$\sum_{j=2}^{\infty} p_j = 1 - p_1.$$

We want to renormalize ($p'_j = \eta p_j$) such that:

$$\sum_{j=2}^{\infty} p'_j = \eta \sum_{j=2}^{\infty} p_j = 1,$$

Thus $\boxed{\eta = 1/(1 - p_1).}$ For the Whois data, $p_1 = 0.0573.$ and $\eta = 1.0608$.

The **complementary cumulative distribution function** (ccdf) for the renormalized probabilities:

$$\mathrm{ccdf}'(\mathrm{x}) = \sum_{\mathrm{j=x}}^{\infty} \mathrm{p}'_{\mathrm{j}} = \eta \sum_{\mathrm{j=x}}^{\infty} \mathrm{p_j} = \eta \ \mathrm{ccdf}(\mathrm{x}).$$

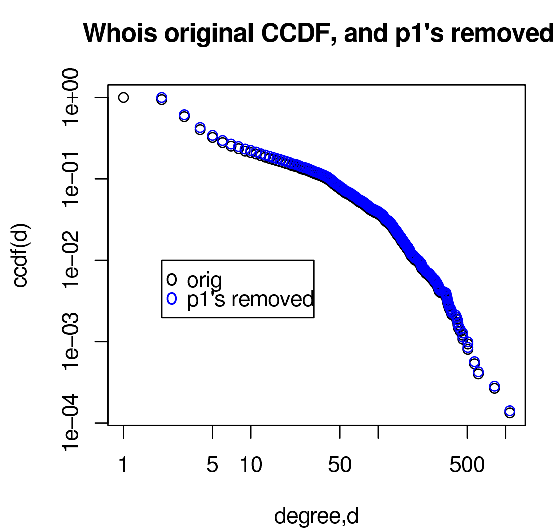

FIG. 1: Original CCDF of Whois data, and the renormalized CCDF'(x) = η CCDF(x).

IV. FITTING TPA TO WHOIS WITH $d \geq 2$

Whois $d \geq 2$ distribution discussed above. TPA with $d \geq 2$ is the same as with $d \geq 1$ except the value of p_{A_2} is defined as in Eqn. (14), in terms of $d = 2$ instead of $d = 1$.

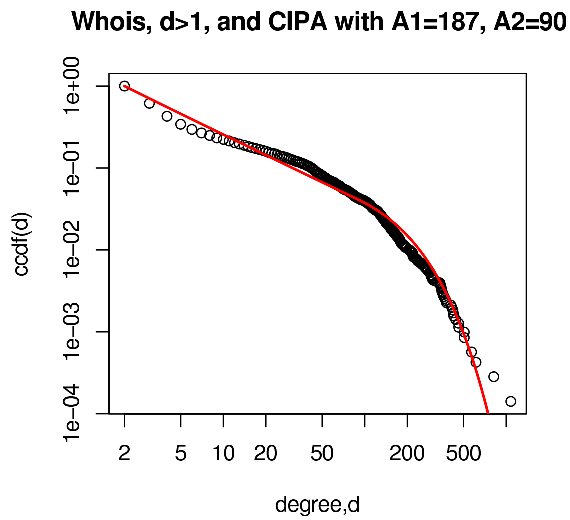

FIG. 2: Whois CCDF for $d \geq 2$. Data points are from the Whois tables. The solid line is the fit to TPA for $d \geq 2$ with $A_1 = 187$ and $A_2 = 90$ (and thus $\gamma = 1.83$). With this fit, $R = 0.986$, thus $R^2 = 0.972$.

V. FITTING PLED TO WHOIS WITH $d \geq 2$

Assuming a PLED: $p(x) = Ax^{-b} \exp(-x/c)$. The normalization constant, A, is determined by the relation:

$$\sum_{x=2}^{\infty} p(x) = 1 = A \sum_{x=2}^{\infty} x^{-b} \exp(-x/c).$$

Then the ccdf:

$$\text{ccdf}(x) = A \sum_{j=x}^{\infty} x^{-b} \exp(-x/c).$$

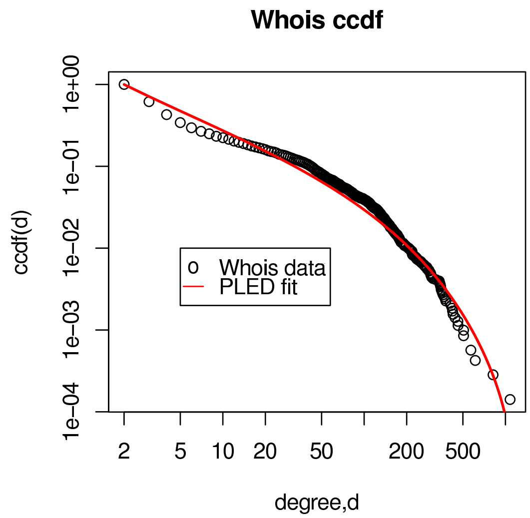

FIG. 3: Whois CCDF for $d \geq 2$. Data points are from the Whois tables. The solid line is the fit $\text{ccdf}(x) = A\sum_{j=x}^{\infty} x^{-b}\exp(-x/c)$, where $b = 1.63$ and $c = 350$. With this fit, $R = 0.985$, thus $R^2 = 0.970$.

[1] R. M. D'Souza, C. Borgs, J. T. Chayes, N. Berger and R. D. Kleinberg. Emergence of Tempered Preferential Attachment From Optimization, to appear *Proc. Natn. Acad. Sci. USA*, 2007. This article will be "open access", and hence freely available for download from the publisher when it appears.

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[4] N. Berger, C. Borgs, J. T. Chayes, R. M. D'Souza, and R. D. Kleinberg. Degree distribution of competition-induced preferential attachment graphs. *Combinatorics, Probability and Computing*, 14(5-6):697–721, 2005.