

MURI Update

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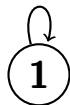
May 12th, 2017

Undirected Preferential Attachment Model

Notations:

- $G(n)$:= the random graph after n -steps.
- $V(n) := \{1, 2, \dots, n\}$, set of nodes in $G(n)$.
- $D_i(n)$:= Degree of node $i \in V_n = \{1, 2, \dots, n\}$.
- $\delta > -1$, parameter.

Initialize with a single node having a self loop.

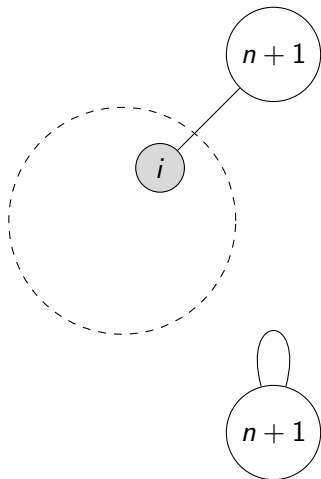


This node is considered as having degree 2, i.e.

$$D_1(1) = 2.$$

From $G(n)$ to $G(n+1)$, consider two scenarios:

Assume **linear** preferential attachment function: $f(i) = i + \delta$.



- (i) The new node $n+1$ attaches to node $i \in V_n$ with probability

$$\frac{D_i(n) + \delta}{(2 + \delta)n + 1 + \delta},$$

and $D_{n+1}(n+1) = 1$.

- (ii) The new node $n+1$ is born with a self loop and this happens with probability

$$\frac{1 + \delta}{(2 + \delta)n + 1 + \delta},$$

and $D_{n+1}(n+1) = 2$.

What Is Known

Define $N_i(n) := \sum_{j=1}^n \mathbf{1}_{\{D_j(n)=i\}}$, then

- (1) The degree counts converge to some deterministic limit:
As $n \rightarrow \infty$, $N_i(n)/n \rightarrow p_i$ with

$$p_i \sim C(\delta) i^{-3-\delta} \quad \text{for } i \rightarrow \infty.$$

- (2) The degree sequence converges to some positive random variable:

$$\frac{D_i(n)}{n^{1/(2+\delta)}} \xrightarrow{\text{a.s.}} \xi_i.$$

- (3) The maximum of the degree sequence also converges:

$$\max_{1 \leq i \leq n} \frac{D_i(n)}{n^{1/(2+\delta)}} \xrightarrow{\text{a.s.}} \max_{i \geq 1} \xi_i < \infty.$$

New Approach

Traditional approach: Use martingale convergence theorem.

Drawbacks:

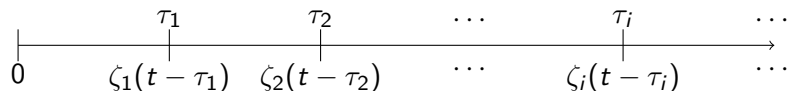
- Not much information on the limit quantity ξ_i has been provided.
- The traditional approach relies heavily on the preferential attachment function.
- For models with sub-linear preferential attachment functions, e.g. $f(i) = (i + \delta)^{-a}$, what will be the corresponding convergence results as in (1)(2)(3)??
- Direct calculations/traditional approaches do not provide much help.

Goal: Find a unified approach such that it can accommodate different types of preferential attachment assumptions.

Tentative solution: Model the degree growth of each single node as a birth process with immigration and find convergence results from the branching process.

Birth Process with Immigration

Let $0 < \tau_1 < \tau_2 < \dots$ be points of a homogeneous Poisson process of rate λ , then we initiate independent linear birth processes at τ_1, τ_2, \dots as below: (cf. Tavaré(1987))



where $\zeta_i(\cdot)$ are independent linear birth processes with transition rate

$$q_{j,j+1} = j, \quad j = 1, 2, \dots$$

Then

$$Bl(t) = \sum_{i=1}^{\infty} \zeta_i(t - \tau_i) \mathbf{1}_{\{t \geq \tau_i\}}$$

is a birth process with immigration (Bl process) and the transition rate is

$$q_{j,j+1} = j + \lambda.$$

Model Construction

We formulate the linear preferential attachment model as below:

$T_1 = 0$
 $BI_1(0) = 2$
 $BI_2(0) = 1$

$T_2 = T_1 + m_2$
 $\sum_{i=1}^2 BI_i(T_2) = 4$

t

- At $T_1 = 0$, initiate two independent BI processes such that $BI_1(0) = 2$, $BI_2(0) = 1$ and transition rate $q_{j,j+1} = j + \delta$.
- Each process has an exponential clock which determines the time of their next jump:
 $\tau_1^{(1)} \sim \text{Expo}(2 + \delta)$, $\tau_2^{(1)} \sim \text{Expo}(1 + \delta)$, and $\tau_1^{(1)} \perp \tau_2^{(1)}$.
- Define $m_2 := \min\{\tau_1^{(1)}, \tau_2^{(1)}\}$, so $m_2 \sim \text{Expo}(3 + 2\delta)$.



- At $T_2 = T_1 + m_2$, one of Bl_1 and Bl_2 jumps so that the sum increases to 4.
- At the same time, initiate a new, independent BI process with $Bl_3(0) = 1$ and transition rate $q_{j,j+1} = j + \delta$.

The construction process then goes on following this dynamic:
Given $G(n)$, we have (with the convention that $T_0 = T_1 = 0$)

$$\tau_i^{(n)} \sim \text{Expo}(Bl_i(T_n - T_{i-1}) + \delta), \quad \tau_{n+1}^{(n)} \sim \text{Expo}(1 + \delta),$$

$$m_n = \bigwedge_{i=1}^{n+1} \tau_i^{(n)} \sim \text{Expo} \left(\sum_{i=1}^{n+1} (Bl_i(T_n - T_{i-1}) + \delta) \right),$$

$$T_n = \sum_{i=2}^n m_i, \quad \{m_i, i \geq 2\} \text{ are all independent from each other.}$$

$T_1 = 0$	$T_2 = T_1 + m_2$	$T_3 = T_2 + m_3$	\dots
$Bl_1(0) = 2$	$\sum_{i=1}^2 Bl_i(T_2) = 4$	$\sum_{i=1}^3 Bl_i(T_3 - T_{i-1}) = 6$	$\dots t$
$Bl_2(0) = 1$	$Bl_3(0) = 1$	$Bl_4(0) = 1$	

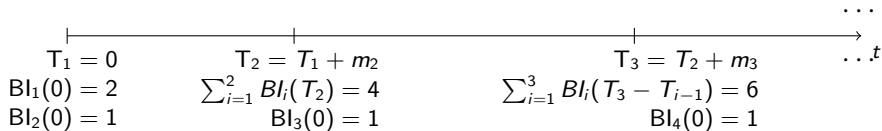
Define the degree sequence as

$$D_i(n) := \begin{cases} Bl_i(T_n - T_{i-1}), & 1 \leq i \leq n \\ 0, & \text{otherwise.} \end{cases}$$

Check:

- $\sum_{i=1}^n D_i(n) = 2n.$
- Attachment probability:
 1. Node $i \in V_n$ is chosen:

$$\begin{aligned} \mathbb{P} \left(\tau_i^{(n)} < \bigwedge_{1 \leq k \leq n+1, k \neq i} \tau_k^{(n)} \middle| G(n) \right) &= \frac{Bl_i(T_n - T_{i-1}) + \delta}{\sum_{i=1}^{n+1} (Bl_i(T_n - T_{i-1}) + \delta)} \\ &= \frac{D_i(n) + \delta}{(2 + \delta)n + 1 + \delta}. \end{aligned}$$



2. Node $n + 1$ is born with a self loop:

$$\mathbb{P} \left(\tau_{n+1}^{(n)} < \bigwedge_{1 \leq k \leq n} \tau_k^{(n)} \mid G(n) \right) = \frac{1 + \delta}{\sum_{i=1}^{n+1} (Bl_i(T_n - T_{i-1}) + \delta)}$$

$$= \frac{1 + \delta}{(2 + \delta)n + 1 + \delta}$$

Asymptotics

Define the counting process

$$N(t) := \frac{1}{2} \sum_{i=1}^{\infty} Bl_i(t - T_{i-1}) \mathbf{1}_{\{t \geq T_i\}} - 1,$$

then it is a birth process with rate

$$q_{j,j+1} = (2 + \delta)(j + 1) + 1 + \delta,$$

and $N(0) = 0$.

Applying the result in Tavaré (1987) gives

$$e^{-(2+\delta)t} N(t) \xrightarrow{a.s.} \text{Gamma} \left(\frac{3 + 2\delta}{2 + \delta}, 1 \right) =: W,$$

so

$$n^{1/(2+\delta)} / e^{T_n} \xrightarrow{a.s.} W^{1/(2+\delta)}.$$

For each $i \geq 2$,

$$e^{-t} B l_i(t) \xrightarrow{a.s.} \text{Gamma}(1 + \delta, 1) =: \sigma_i, \quad t \rightarrow \infty.$$

Then

$$\frac{D_i(n)}{e^{T_n - T_{i-1}}} = \frac{B l_i(T_n - T_{i-1})}{e^{T_n - T_{i-1}}} \xrightarrow{a.s.} \sigma_i, \quad n \rightarrow \infty.$$

Hence,

$$\frac{D_i(n)}{e^{T_n}} \xrightarrow{a.s.} \sigma_i e^{-T_{i-1}}, \quad n \rightarrow \infty,$$

and $\sigma_i \perp T_{i-1}$.

Recall that $n^{1/(2+\delta)}/e^{T_n} \xrightarrow{a.s.} W^{1/(2+\delta)}$. Then it follows

$$\frac{D_i(n)}{n^{1/(2+\delta)}} \xrightarrow{a.s.} W^{-1/(2+\delta)} \sigma_i e^{-T_{i-1}}, \quad n \rightarrow \infty.$$

Can also show that

$$\bigvee_{i=1}^n \frac{D_i(n)}{n^{1/(2+\delta)}} \xrightarrow{\text{a.s.}} W^{-1/(2+\delta)} \bigvee_{i \geq 1} \sigma_i e^{-T_{i-1}}, \quad n \rightarrow \infty,$$

and

$$\sum_{i=1}^n \epsilon_{D_i(n)/n^{1/(2+\delta)}} \Rightarrow \sum_{i=1}^{\infty} \epsilon_{W^{-1/(2+\delta)} \sigma_i e^{-T_{i-1}}}$$

in $M_+(0, \infty]$.

$$\frac{D_i(n)}{n^{1/(2+\delta)}} \xrightarrow{a.s.} W^{-1/(2+\delta)} \sigma_i e^{-T_{i-1}}, \quad n \rightarrow \infty.$$

Question:

- (i) When $\delta = 0$, there is a representation in Bollobás (2003):

$$\frac{D_i(n)}{2n^{1/2}} \xrightarrow{a.s.} (\text{Gamma}(i, 1))^{1/2} - (\text{Gamma}(i-1, 1))^{1/2}.$$

How is it related to our result?

- (ii) What does this tell about consistency of the Hill estimator?
- (iii) Can we generalize this formulation to accommodate the directed linear PA model?
(Multi-type branching process??)
- (iv) When the PA function is changed, how will it affect the asymptotic results?
(Will lead to a modification on the transition rate of the birth process.)