

About the local clustering coefficient in preferential attachment graphs

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Plan

- 1 Models based on the preferential attachment
- 2 Theoretical analysis of the general case and the local clustering coefficient
- 3 Conclusion

Degree distribution

Real-world networks often have the power law degree distribution:

$$\frac{\#\{v : \deg(v) = d\}}{n} \approx \frac{c}{d^\gamma},$$

where $2 < \gamma < 3$.

Clustering coefficient

Global clustering coefficient of a graph G :

$$C_1(n) = \frac{3\#(\text{triangles in } G)}{\#(\text{pairs of adjacent edges in } G)}.$$

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Average local clustering coefficient

- T^i is the number of edges between the neighbors of a vertex i
- P_2^i is the number of pairs of neighbors
- $C(i) = \frac{T^i}{P_2^i}$ is the local clustering coefficient for a vertex i
- $C_2(n) = \frac{1}{n} \sum_{i=1}^n C(i)$ – average local clustering coefficient

Preferential attachment

Idea of preferential attachment [Barabási, Albert]:

- Start with a small graph
- At every step we add new vertex with m edges
- The probability that a new vertex will be connected to a vertex i is proportional to the degree of i

PA-class of models

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$$A \frac{\deg(i)}{n} + B \frac{1}{n} + O\left(\frac{(\deg(i))^2}{n^2}\right)$$

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- $2mA + B = m$, $0 \leq A \leq 1$

T -subclass

Triangles property:

The probability that the degree of two vertices i and j increases by one equals

$$e_{ij} \frac{D}{mn} + O\left(\frac{d_i^n d_j^n}{n^2}\right)$$

Here e_{ij} is the number of edges between vertices i and j in G_m^n and D is a positive constant.

Bollobás–Riordan, Buckley–Osthus, Móri, etc.

- Fix some positive number a – "initial attractiveness". (Bollobás–Riordan model: $a = 1$).
- Start with a graph with one vertex and m loops.
- At n -th step add one vertex with m edges.
- We add m edges one by one. The probability to add an edge $n \rightarrow i$ at each step is proportional to $\text{indeg}(i) + am$.

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- **Outdegree:** m
- **Triangles property:** $D = 0$
- **PA-condition:** $A = \frac{1}{1+a}$
- **Degree distribution:** Power law with $\gamma = 2 + a$
- **Global clustering:** $\frac{(\log n)^2}{n}$ ($a = 1$), $\frac{\log n}{n}$ ($a > 1$)

Degree distribution

Let $N_n(d)$ be the number of vertices with degree d in G_m^n .

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Expectation

For every $d \geq m$ we have

$$\mathbb{E}N_n(d) = c(m, d) \left(n + O\left(d^{2+\frac{1}{A}}\right) \right),$$

where

$$c(m, d) = \frac{\Gamma\left(d + \frac{B}{A}\right) \Gamma\left(m + \frac{B+1}{A}\right)}{A \Gamma\left(d + \frac{B+A+1}{A}\right) \Gamma\left(m + \frac{B}{A}\right)} \sim \frac{\Gamma\left(m + \frac{B+1}{A}\right) d^{-1-\frac{1}{A}}}{A \Gamma\left(m + \frac{B}{A}\right)}$$

and $\Gamma(x)$ is the gamma function.

Degree distribution

Concentration

For every $d = d(n)$ we have

$$\mathbb{P}(|N_n(d) - \mathbb{E}N_n(d)| \geq d\sqrt{n} \log n) = O(n^{-\log n}).$$

Therefore, for any $\delta > 0$ there exists a function $\varphi(n) = o(1)$ such that

$$\lim_{n \rightarrow \infty} \mathbb{P}\left(\exists d \leq n^{\frac{A-\delta}{4A+2}} : |N_n(d) - \mathbb{E}N_n(d)| \geq \varphi(n) \mathbb{E}N_n(d)\right) = 0.$$

Global clustering

Let $P_2(n)$ be the number of all path of length 2 in G_m^n .

$P_2(n)$

- (1) If $2A < 1$, then **whp** $P_2(n) \sim \left(2m(A + B) + \frac{m(m-1)}{2}\right) \frac{n}{1-2A}$.
- (2) If $2A = 1$, then **whp** $P_2(n) \propto n \log(n)$.
- (3) If $2A > 1$, then **whp** $P_2(n) \propto n^{2A}$.

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Triangles

Whp the number of triangles $T(n) \sim D n$.

Global clustering

Global clustering

- (1) If $2A < 1$ then **whp** $C_1(n) \sim \frac{3(1-2A)D}{(2m(A+B) + \frac{m(m-1)}{2})}$.
- (2) If $2A = 1$ then **whp** $C_1(n) \propto (\log n)^{-1}$.
- (2) If $2A > 1$ then **whp** $C_1(n) \propto n^{1-2A}$.

Average local clustering

Average local clustering

Whp

$$C_2(n) \geq \frac{1}{n} \sum_{i:\deg(i)=m} C(i) \geq \frac{2cD}{m(m+1)}.$$

Local clustering: definition

Let $T_n(d)$ be the number of triangles on the vertices of degree d in G_m^n .

Let $N_n(d)$ be the number of vertices of degree d in G_m^n .

Local clustering coefficient over vertices of degree d :

$$C(d) = \frac{T_n(d)}{N_n(d) \frac{d(d-1)}{2}}$$

Local clustering: number of triangles

Let $T_n(d)$ be the number of triangles on the vertices of degree d in G_m^n . Then

Expectation

If $2A < 1$ then

$$ET_n(d) = K(d) \left(n + \theta \left(C \cdot d^{2+\frac{1}{A}} \right) \right),$$

If $2A = 1$ then

$$ET_n(d) = K(d) \left(n + \theta \left(C \cdot d^{2+\frac{1}{A}} \cdot \log(n) \right) \right),$$

If $2A > 1$ then

$$ET_n(d) = K(d) \left(n + \theta \left(C \cdot d^{2+\frac{1}{A}} \cdot n^{2A-1} \right) \right),$$

where

$$K(d) = \left[\frac{\sum_{i=1}^d \frac{D(i-1)}{m[A(i-1)+B]}}{\sum_{i=1}^m \frac{(i-1)}{m[A(i-1)+B]}} \right] \cdot c(m, d) d \rightarrow \infty \frac{D}{A \cdot \sum_{i=1}^m \frac{(i-1)}{[A(i-1)+B]}} \cdot \frac{\Gamma(m + \frac{B+1}{A})}{A \Gamma(m + \frac{B}{A})} \cdot d^{-\frac{1}{A}}$$

and $\theta(X)$ is some function such that $|\theta(X)| < X$.

Idea of the proof

- The following recurrent formula holds:

$$\begin{aligned} ET_{n+1}(d) &= ET_n(d) - ET_n(d) \cdot \left[\frac{Ad + B}{n} + O\left(\frac{d^2}{n^2}\right) \right] + \\ &+ ET_n(d-1) \cdot \left[\frac{A(d-1) + B}{n} + O\left(\frac{(d-1)^2}{n^2}\right) \right] + \\ &+ \sum_{\substack{j:j \text{ is a neighbor} \\ \text{of a vertex of degree } d}} (d-1)N_n(d-1) \left[\frac{D}{mn} + O\left(\frac{(d-1) \cdot d_j}{n^2}\right) \right] \end{aligned}$$

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- Next we prove that $ET_n(d) = K(d)(n + \Theta(\dots))$, where $K(d) = K(d-1) \frac{Ad-A+B}{Ad+B+1} + c(m, d-1) \frac{D(d-1)}{m(Ad+B+1)}$

Number of triangles

Let G_m^n belong to T-subclass of PA-class. Then

Concentration

If $2A < 1$ then

$$P(|T_n(d) - ET_n(d)| \geq d^2 \sqrt{n} \log n) = O(n^{-\log n}),$$

Therefore, for any $\delta > 0$ there exists a function $\varphi(n) = o(1)$ such that

$$\lim_{n \rightarrow \infty} P\left(\exists d \leq n^{\frac{A-\delta}{4A+2}} : |T_n(d) - ET_n(d)| \geq \varphi(n) ET_n(d)\right) = 0.$$

Number of triangles

Concentration

If $2A = 1$ then

$$\mathbb{P}(|T_n(d) - \mathbb{E}T_n(d)| \geq d^2 \sqrt{n} \log^2 n) = O(n^{-\log n}),$$

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Number of triangles

Concentration

If $1 < 2A < \frac{3}{2}$ then

$$\mathbb{P} \left(|T_n(d) - \mathbb{E}T_n(d)| \geq d^2 n^{2A - \frac{1}{2}} \log n \right) = O(n^{-\log n}),$$

Therefore, for any $\delta > 0$ there exists a function $\varphi(n) = o(1)$ such that

$$\lim_{n \rightarrow \infty} \mathbb{P} \left(\exists d \leq n^{\frac{A(3-4A)-\delta}{4A+2}} : |T_n(d) - \mathbb{E}T_n(d)| \geq \varphi(n) \mathbb{E}T_n(d) \right) = 0.$$

Idea of the proof

Azuma, Hoeffding

Let $(X_i)_{i=0}^n$ be a martingale such that $|X_i - X_{i-1}| \leq c_i$ for any $1 \leq i \leq n$. Then

$$P(|X_n - X_0| \geq x) \leq 2e^{-\frac{x^2}{2\sum_{i=1}^n c_i^2}}$$

for any $x > 0$.

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for any $x > 0$.

- $X_i(d) = E(T_n(d) \mid G_m^i)$, $i = 0, \dots, n$.
- Note that $X_0(d) = ET_n(d)$ and $X_n(d) = T_n(d)$.
- $X_n(d)$ is a martingale.
- For any $i = 0, \dots, n-1$: $|X_{i+1}(d) - X_i(d)| \leq Md^2$ (for the case $2A < 1$), where $M > 0$ is some constant.

Idea of the proof

Fix $0 \leq i \leq n - 1$ and some graph G_m^i .

$$\begin{aligned} & \left| \mathbb{E} \left(T_n(d) \mid G_m^{i+1} \right) - \mathbb{E} \left(T_n(d) \mid G_m^i \right) \right| \leq \\ & \leq \max_{\tilde{G}_m^{i+1} \supset G_m^i} \left\{ \mathbb{E} \left(T_n(d) \mid \tilde{G}_m^{i+1} \right) \right\} - \min_{\tilde{G}_m^{i+1} \supset G_m^i} \left\{ \mathbb{E} \left(T_n(d) \mid \tilde{G}_m^{i+1} \right) \right\}. \end{aligned}$$

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$$\hat{G}_m^{i+1} = \arg \max \mathbb{E} \left(T_n(d) \mid \tilde{G}_m^{i+1} \right), \quad \bar{G}_m^{i+1} = \arg \min \mathbb{E} \left(T_n(d) \mid \tilde{G}_m^{i+1} \right).$$

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For $i + 1 \leq t \leq n$ put $\delta_t^i(d) = \mathbb{E} \left(T_t(d) \mid \hat{G}_m^{i+1} \right) - \mathbb{E} \left(T_t(d) \mid \bar{G}_m^{i+1} \right)$.

Idea of the proof

Fix $0 \leq i \leq n - 1$ and some graph G_m^i .

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$$\begin{aligned} \delta_{t+1}^i(d) & \leq \delta_t^i(d) \left(1 - \frac{Ad + B}{t} + O\left(\frac{d^2}{t^2}\right) \right) + \\ & + \delta_t^i(d - 1) \left(\frac{A(d - 1) + B}{t} + O\left(\frac{d^2}{t^2}\right) \right) + O\left(\frac{d^2}{t}\right). \end{aligned}$$

Local clustering

Let G_m^n belong to T-subclass of PA-class. Then

Local clustering coefficient

If $2A \leq 1$ then for any $\delta > 0$ there exists a function $\varphi(n) = o(1)$ such that

$$\lim_{n \rightarrow \infty} \mathbb{P} \left(\exists d \leq n^{\frac{A-\delta}{4A+2}} : \left| C(d) - \frac{2D}{A \cdot \sum_{i=1}^m \frac{(i-1)}{[A(i-1)+B]}} \cdot \frac{1}{d} \right| \geq \varphi(n) \frac{1}{d} \right) = 0$$

If $1 < 2A < \frac{3}{2}$ then for any $\delta > 0$ there exists a function $\varphi(n) = o(1)$ such that

$$\lim_{n \rightarrow \infty} \mathbb{P} \left(\exists d \leq n^{\frac{A(3-4A)-\delta}{4A+2}} : \left| C(d) - \frac{2D}{A \cdot \sum_{i=1}^m \frac{(i-1)}{[A(i-1)+B]}} \cdot \frac{1}{d} \right| \geq \varphi(n) \frac{1}{d} \right) = 0$$

Generalized preferential attachment models:

- Power law degree distribution with any exponent $\gamma > 2$
- Constant global clustering coefficient **only** for $\gamma > 3$
- Constant average local clustering coefficient
- Local clustering $C(d) \sim \frac{C}{d}$ for $\gamma > \frac{7}{3}$

Thank You!

Questions?