

Preferential attachment model

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①

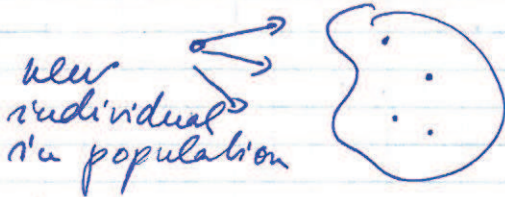
ER Random graph
 Generalized random graph
 Config. model } Static models, size of graph fixed

Dynamic models → lead to power law degree sequences (scale-free nature)

- Vertices added sequentially, ~~connected then~~ with a number of edges

connected proportionally to the degree of the receiving vertex

(favoring v with $\text{deg } v$ large)



ER: new edge with equal prob.

bias towards more social individuals

(hubs, high degree nodes in network)

Barabási & Albert (Albert-László Barabási) & Réka

(Yule ~ group model with evolution of species)

Simon ~ occurrence of words in large text

① prob. ($k+1$)th word already appeared i times \sim # words scanned

② constant prob. ($k+1$)th word is new i times frequency distr. \sim power law

$t_0=0$: no vertices

$\forall t$: 1 new vertex with m stubs linking to m different vertices

$$P(v \text{ connected to } v_{old}) \sim$$

$$\sim \text{deg}(v_{old})$$

rigorous

Bollobás, Riordan, Spencer & Tuzi

lack of formal def.

1, how to start ($\text{deg } v_i = 0$ initially)

2, $\sum_i P(v \sim v_{old}) = 1$ expected degree = 1 not m .

3, giving marginals distribution you can get anything.

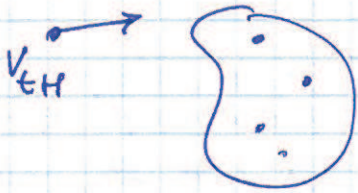
Model

Graph sequence $\{PA_t(m, \delta)\}_{t=1}^{\infty}$ $\{PA_{(m, \delta)}(t)\}_{t=1}^{\infty}$ PA_t : graph t vertices mt edges

1, $m=1$
 $t=1$ \rightarrow 1 vertex, selfloop

vertices $\{v_{11}, v_{21}, \dots, v_t\}$ $\text{deg } v_i \rightarrow D_i(t)$ self-loop = 2

$PA_{tH}(1, \delta) \mid PA_t(1, \delta)$ is given by



$$P(v_{tH} \leftrightarrow v_{tH}) = \frac{1+\delta}{t \cdot (2+\delta) + (1+\delta)}$$

$$P(v_{tH} \leftrightarrow v_i) = \frac{D_i(t) + \delta}{t(2+\delta) + (1+\delta)} = p_i$$

$\delta > -1$ parameter

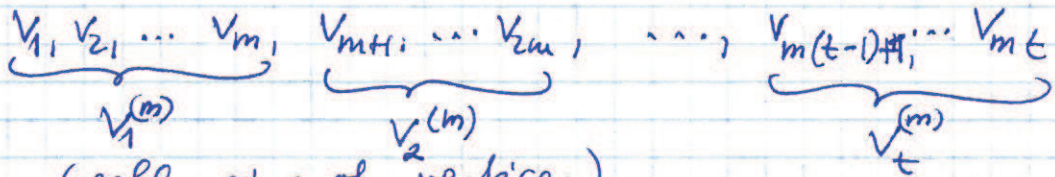
Ex. 1. $D_i(t) \geq 1 \quad \forall t \geq i$

$$\Rightarrow D_i(t) + \delta \geq 0 \quad \forall \delta \geq -1$$

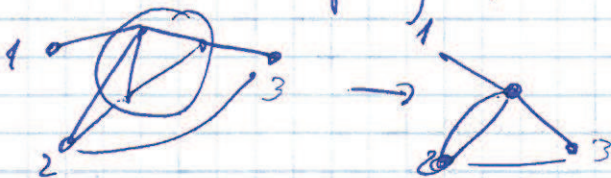
Ex. 2. $\sum_{i=1}^{t+1} p_i = 1$ (prove) $\frac{(1+\delta) + \sum (D_i(t) + \delta)}{t(2+\delta) + (1+\delta)} =$

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

2, $m > 1$. Consider $PA_{mt}(1, \frac{\delta}{m})$



identify vertices (collapsing of vertices)



multigraph with t vertices, mt edges
 $\sum D_i(t) = 2mt$

$$PA_{mt}(1, \frac{\delta}{m})$$

$$P(v \sim v^*) \sim \frac{\text{deg } v + \frac{\delta}{m}}{m}$$

$$P(v \sim v_1, v_2, \dots, v_m) \sim \frac{\sum_{i=1}^m \text{deg } v_i + \delta}{m}$$

\Rightarrow In this model

Degrees updated after each stub is connected

$v_{u(t-1)+1}, v_{u(t-1)+2}, \dots, v_{ut}$

Intermediate updating

② Barabási-Albert model $\sigma = 0$

$PA_t(m, \sigma)$ constructed via $PA_{mt}(1, \frac{\sigma}{m})$

immediate construction:

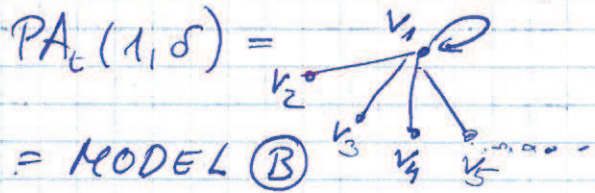
$PA_1(m, \sigma)$ $t=1$ \rightarrow m loops

$PA_{t+1}(m, \sigma)$ add m stubs $\left\{ \begin{array}{l} v_{t+1} \\ \vdots \\ v_j \end{array} \right\}$ added sequentially, intermediate update of degrees

$P(v_{t+1}, j\text{th stub} \sim v_i) \sim D_i(j-1, t) + \sigma$

$D_i(j, t)$ = degree of v_i after j th stub added to the graph.

Ex. 3. Prove, $\sigma = -1$, $m > 1$?



③ Variation NO SELF LOOP = MODEL (B)

$PA_t^{(B)}(m, \sigma)$

$t=2$ $v_1 \rightarrow v_2$

growth rule $P(v_{t+1} \rightarrow v_i | PA_t^{(B)}(1, \sigma)) = \frac{D_i(t) + \sigma}{t(2+\sigma)}$

Advantage:

- no self loop
- connected graph

$m \geq 2, \sigma > -m$ $PA_{mt}^{(B)}(1, \frac{\sigma}{m})$

$P_t(\text{loop})$ very small if t large

④ Remark $\sigma = 0 \rightarrow \sigma = \infty$

interpolation between BA & ER

$m=1$

$\alpha \in [0, 1]$ draw n i.i.d. $X_{t+1} = \begin{cases} 0 & \alpha \\ 1 & 1-\alpha \end{cases}$

$X_{t+1} = 0 \rightarrow$ apply uniform attachment

$= 1 \rightarrow$ pref. attachment $P(v_{t+1} \rightarrow v_i) = \frac{D_i(t)}{2t}$

$PA_t^{B'}(1, \alpha)$ $\alpha > 0$

Prove that $\alpha = \frac{\sigma}{2+\sigma}$; $\left\{ PA_t^{B'}(1, \alpha) \right\}_{t=2}^{\infty} \sim \left\{ PA_t^B(1, \sigma) \right\}_{t=1}^{\infty}$

⑤ $m=1$ Prove $D_i(t) \rightarrow \infty$ a.s.

$I_t := \text{BER}\left(\frac{1+\sigma}{t(2+\sigma)+1+\sigma}\right) \sum_{s=0}^t I_s \leq D_i(t) \rightarrow \infty$ a.s.

8.2. Degrees of ^{fixed} vertices

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$$\Gamma(t) = \int_0^\infty x^{t-1} e^{-x} dx; \quad \Gamma(t+1) = t\Gamma(t)$$

$$\frac{\Gamma(t+a)}{\Gamma(t)} = t^a (1 + O(\frac{1}{t}))$$

Thm. 8.1. $m=1; \delta > -1.$

(Stirling formula)

$$\frac{D_i(t)}{t^{\frac{1}{2+\delta}}} \longrightarrow \xi_i \quad \text{a.s. as } t \rightarrow \infty$$

$$e^{-t} t^{\frac{t+1}{2}} \leq \Gamma(t+1) \leq \frac{1}{e^{t/2}}$$

$$\mathbb{E}(D_i(t) + \delta) = (1+\delta) \cdot \frac{\Gamma(t+1) \Gamma(i - \frac{1}{2+\delta})}{\Gamma(t + \frac{1+\delta}{2+\delta}) \cdot \Gamma(i)}$$

Proof. Via martingales.

$$\mathbb{E}(D_i(t+1) + \delta | D_i(t)) = D_i(t) + \delta + \mathbb{E}(D_i(t+1) - D_i(t) | D_i(t)) = D_i(t) + \delta + \underbrace{0}_{\text{v.l.}}$$

$$= (D_i(t) + \delta) \cdot \frac{(t+1)(2+\delta)}{(2+\delta)t + 2 + \delta} \cdot \frac{D_i(t) + \delta}{(2+\delta)t + (1+\delta)} =$$

initializing

$$\mathbb{E}(D_i(i) + \delta) = (1+\delta) + \frac{1+\delta}{(2+\delta)(i-1) + 1 + \delta} = \frac{(1+\delta)(2+\delta) i^0}{(2+\delta)(i-1) + (1+\delta)} = (1+\delta) c_{i-1}$$

we can get a martingale $M_i(t)$

$$M_i(t) = \frac{D_i(t) + \delta}{1+\delta} \cdot \frac{t+1}{s} \cdot \frac{1}{c_s} \quad \mathbb{E} M_i(t) = 1$$

$$\mathbb{E}(M_i(t+1) | M_i(t)) = \mathbb{E}(D_i(t+1) + \delta) \cdot \frac{1}{1+\delta} \cdot \frac{t+2}{s+1} \cdot \frac{1}{c_{s+1}} = c_t \cdot (D_i(t) + \delta) \cdot \frac{1}{1+\delta} \cdot \frac{t+1}{s} \cdot \frac{1}{c_s} = (D_i(t) + \delta) \cdot \frac{t+1}{s} \cdot \frac{1}{c_s} = M_i(t)$$

$$\mathbb{E} M_i(i) = \frac{\mathbb{E}(D_i(i) + \delta)}{1+\delta} \cdot \frac{1}{c_{i-1}} = 1 \checkmark$$

$$\frac{1}{1+\delta} = M_i(t) \checkmark$$

martingale convergence theorem

2.21.

M_n be a martingale wrt \mathcal{X}_n

$$\mathbb{E}|M_n| \leq B \quad \forall n \geq 0 \Rightarrow M_n \xrightarrow{\text{a.s.}} M_\infty$$

for some limiting random variable M_∞ which

$$\mathbb{P}(M_\infty < \infty) = 1 \quad (\text{if } \mathbb{E} M_n^2 < M \quad \forall n \Rightarrow \lim_{n \rightarrow \infty} \mathbb{E} M_n^2 = \mathbb{E} M_\infty^2)$$

$$\begin{aligned}
 \text{8. } \prod_{s=i-1}^{t-1} \frac{1}{c_s} &= \prod_{s=i-1}^{t-1} \frac{t-1}{(2+\delta)s + (1+\delta)} = \prod_{s=i-1}^{t-1} \frac{s + \frac{1+\delta}{2+\delta}}{s+1} = \\
 &= \frac{\Gamma(t + \frac{1+\delta}{2+\delta})}{\Gamma(i - \frac{1}{2+\delta})} \cdot \frac{1}{\frac{\Gamma(t+1)}{\Gamma(i)}} = \frac{\Gamma(i)}{\Gamma(i - \frac{1}{2+\delta})} \cdot \frac{\Gamma(t + \frac{1+\delta}{2+\delta})}{\Gamma(t+1)} \approx \\
 &\approx i^{-\frac{1}{2+\delta}} \cdot t^{\frac{(1+\delta)}{2+\delta} - 1} \rightarrow t^{-\frac{1}{2+\delta}}
 \end{aligned}$$

$$\frac{D_i(t) + \delta}{1+\delta} \cdot \frac{t^{-\frac{1}{2+\delta}} (1 + \delta \frac{1}{t})}{\Gamma(i)} \rightarrow M_{i,\infty}$$

$\mathbb{E} M_{i,\infty} = 1$

$$\Rightarrow \frac{D_i(t)}{t^{\frac{1}{2+\delta}}} \rightarrow (1+\delta) \frac{\Gamma(i - \frac{1}{2+\delta})}{\Gamma(i)} M_{i,\infty}$$

$$\mathbb{E} \left(\frac{D_i(t) + \delta}{t^{\frac{1}{2+\delta}}} \right) \Rightarrow ? - \mathbb{P}(\xi_i = 0) = 0? \text{ or not?}$$

Result for $m > 1$ $PA_t(m, \delta) \leftrightarrow PA_{ut}(1, \frac{\delta}{m})$

$$\mathbb{E}_{m,\delta} (D_i(t)) = \sum_{s=1}^m \mathbb{E}_{1, \frac{\delta}{m}} (D_{m(i-1)+s}(ut))$$

$$\frac{D_i(t)}{(mt)^{\frac{-1}{2+\frac{\delta}{m}}}} \xrightarrow{\text{a.s.}} \xi_i' \quad \xi_i' = \sum_{j=(i-1)m+1}^{mi} \xi_j \quad \uparrow \xi_j \text{ as above in } PA(1, \frac{\delta}{m})$$

Power law heuristic

$$\mathbb{E} D_i(t) \sim a_m \cdot \left(\frac{t}{i}\right)^{\frac{1}{2+\frac{\delta}{m}}}$$

$$N_{\geq k}(t) = \#\{i: D_i(t) \geq k\}$$

heuristic:

$$\begin{aligned}
 N_{\geq k}(t) &\approx \#\{i: \mathbb{E} D_i(t) \geq k\} \approx \#\{i: a_m \left(\frac{t}{i}\right)^{\frac{1}{2+\frac{\delta}{m}}} \geq k\} \\
 &= \sum_{i=1}^t \mathbb{1}(i < t a_m^{\frac{2+\frac{\delta}{m}}{1}} \cdot k^{-(2+\frac{\delta}{m})}) = t \cdot a_m^{\frac{2+\frac{\delta}{m}}{1}} \cdot k^{\frac{-(2+\frac{\delta}{m})}{1}}
 \end{aligned}$$

$$\frac{N_{\geq k}(t)}{t} \approx k^{-\frac{(2+\frac{\delta}{m})}{1}} \sim \tau^{-1} = 2 + \frac{\delta}{m} \quad \tau = 3 + \frac{\delta}{m}$$

Degree sequence?

$$P_k(t) = \frac{1}{t} \sum_{i=1}^t \mathbb{1}(D_i(t) = k)$$

proportion of vertices with degree k at time t

($m \geq 1, \delta > -m, P_k = 0$ for $k \leq m-1$)

let's define

$$\otimes P_k = \binom{2 + \frac{\delta}{m}}{k} \frac{\Gamma(k + \delta) \Gamma(m + 2 + \delta + \frac{\delta}{m})}{\Gamma(m + \delta) \Gamma(k + 3 + \delta + \frac{\delta}{m})}$$

$m=1,$
 ~~$P_k =$~~

$\delta=0, k \geq m$ $P_k = \frac{\Gamma(k)}{\Gamma(m)} \cdot \frac{\Gamma(m+2)}{\Gamma(k+3)} = \frac{2 \cdot m(m+1)}{k(k+1)(k+2)}$

Aim is to prove this is the limiting distribution

1. $\{P_k\}_{k=1}^{\infty}$ prob. distribution.

$$\frac{\Gamma(k+a)}{\Gamma(k+b)} = \frac{1}{b-a+1} \left(\frac{\Gamma(k+a)}{\Gamma(k-1+b)} - \frac{\Gamma(k+1+a)}{\Gamma(k+b)} \right) \quad \boxed{\Gamma(t+1) = t\Gamma(t)}$$

Proof elem.

$$\frac{1}{b-a+1} \frac{\Gamma(k+a)(k+b) - \Gamma(k+a)(k+a)}{\Gamma(k+b)} = \frac{\Gamma(k+a)(b-a-1)}{\Gamma(k+b)(b-a+1)} = \checkmark$$

$a = \delta \quad b = 3 + \delta + \frac{\delta}{m}$

$$P_k = \frac{\Gamma(m + 2 + \delta + \frac{\delta}{m})}{\Gamma(m + \delta)} \cdot \left(\frac{\Gamma(k + \delta)}{\Gamma(k + 2 + \delta + \frac{\delta}{m})} - \frac{\Gamma(k + 1 + \delta)}{\Gamma(k + 3 + \delta + \frac{\delta}{m})} \right)$$

$2 + \frac{\delta}{m}$ is not

$\Rightarrow \sum_{k \geq 1} P_k = \sum_{k \geq m} P_k = \frac{\Gamma(m + 2 + \delta + \frac{\delta}{m})}{\Gamma(m + \delta)} \cdot \frac{\Gamma(m + \delta)}{\Gamma(m + 2 + \delta + \frac{\delta}{m})} = 1$

telescopic sum first term

+ $|P_k \geq 0|$

Theorem

Degree sequence $\delta > -m, m \geq 1$ fix; $\exists C = C(m, \delta) > 0$

$$\mathbb{P} \left(\max_k |P_k(t) - p_k| \geq C \cdot \sqrt{\frac{\log t}{t}} \right) = o(1) \text{ as } t \rightarrow \infty$$

$$P_k = C_{m, \delta} \cdot k^{-\gamma} \cdot \left(1 + O\left(\frac{1}{k}\right) \right)$$

$\gamma = 3 + \frac{\delta}{m} > 2$

$$C_{m, \delta} = \binom{2 + \frac{\delta}{m}}{m} \frac{\Gamma(m + 2 + \delta + \frac{\delta}{m})}{\Gamma(m + \delta)}$$

Proof of Thm 8.2. In $PA_t(m, \sigma)$ as $\forall C > m\sqrt{\sigma}$, as $t \rightarrow \infty$

$$N_k(t) = \sum_{i=1}^t \mathbb{1}(D_i(t) = k) = t P_k(t)$$

$$P\left(\max_k |P_k(t) - p_k| \geq C \cdot \sqrt{\frac{\log t}{t}}\right) \leq o(1)$$

① $P\left(\max_k |N_k(t) - \mathbb{E}N_k(t)| \geq C \cdot \sqrt{t \log t}\right) = o(1)$

② ~~P~~ $|\mathbb{E}N_k(t) - tp_k| \leq C$

Proof of ①: $N_k(t) = 0$ $k > m(t+1)$

$$P\left(\max_k |N_k(t) - \mathbb{E}N_k(t)| \geq C\sqrt{t \log t}\right) = P\left(\max_{k \leq m(t+1)} |N_k(t) - \mathbb{E}N_k(t)| \geq C\sqrt{t \log t}\right)$$

max $\leq \sum$

$$\leq \sum_{k=1}^{m(t+1)} P(|N_k(t) - \mathbb{E}N_k(t)| \geq C\sqrt{t \log t})$$

conditional do up to time n:

$$M_n = \mathbb{E}(N_k(t) | \mathcal{F}_n) \xleftarrow{PA_n(m, \sigma)} \text{Doob mart.}$$

prove, each term $o(\frac{1}{t})$

1. $N_k(t)$ bounded $\Rightarrow \mathbb{E}N_k(t) \leq t < \infty$ (total number of vertices)

$$\mathbb{E}(M_{n+1} | PA_n) = \mathbb{E}(\mathbb{E}(N_k(t) | PA_{n+1}) | PA_n) = \mathbb{E}(N_k | PA_n) = M_n$$

$$M_0 = \mathbb{E}N_k(t) \quad M_t = N_k(t)$$

$$\Rightarrow |M_t - M_0| \leq \sum_{i=1}^t |M_i - M_{i-1}|$$

Azuma-Koeffizient

$$|M_n - M_{n-1}| \leq 2m \quad \forall n \in [t] \text{ a.s.}$$

Markingales with bounded differences $|x_k - x_{k-1}| < c_k$
 $\Rightarrow \forall N, \forall t > 0$
 $PP(|X_N - X_0| \geq t) \leq e^{-\frac{t^2}{2 \sum c_k^2}}$

$$M_n - M_{n-1} = \sum_{i=1}^t P(D_i(t) = k | PA_n) - P(D_i(t) = k | PA_{n-1}) =$$

~~extra~~ extra information gained in step n

PA_n : info to which the first m s edges are attached

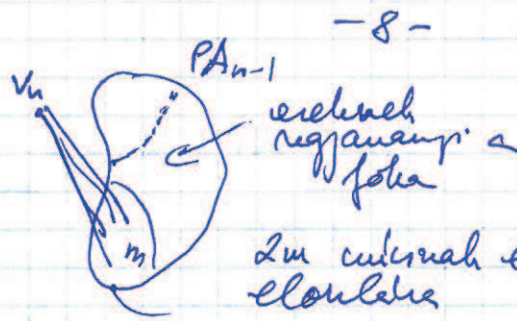
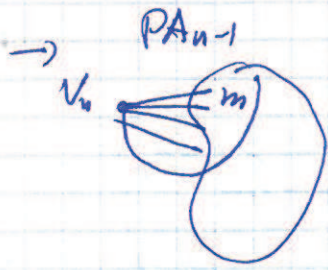


$PA_n: G(v_1, \dots, v_{n-1})$ is fixed + v_n has m possible neighbors
 $m \leq m$ db c_k $\leq m$ \Rightarrow $c_k \leq m$ \Rightarrow $\sum c_k^2 \leq m^2 t$

$P(\dots | PA_n)$ \leftarrow expectation w.r.t. all possible ways of attaching the remaining vertices v_{n+1}, \dots, v_t

eg v_{n-t} \rightarrow legfellebb $2m$ \Rightarrow $c_k \leq m$ \Rightarrow $\sum c_k^2 \leq m^2 t$

$$= P(D_i(t) = k | \dots)$$



forrás
egyszer

A löbből
nem vektorad
(a behatolás
valószínűsége)
1-εN
de csak
ebben a 2m
különbség
van valószínű
sértékek
a löbbi vektorad

Koefficiens: $M_T - M_0$

$$P\left(\sum_{k=1}^t M_k - M_{k-1} \geq a\right) \leq 2 \cdot e^{-\frac{a^2}{2(2m)^2 t}} = 2e^{-\frac{a^2}{8m^2 t}}$$

$$a := C\sqrt{t \log t}$$

$$= 2e^{-\frac{C^2 t \log t}{8m^2 \cdot t}} = 2\left(\frac{1}{t}\right)^{\frac{C^2}{8m^2}}$$

$C^2 > 8m^2$ ez $\frac{1}{t}$

Remark 1. $N_{\geq k}(t) = \sum_{l=k}^{\infty} N_l(t)$ concentrates as well
 Try to prove it by using $M_n' = E(N_{\geq k}(t) | P_{A_n})$
 $P(|N_{\geq k}(t) - E N_{\geq k}(t)| \geq C\sqrt{t \log t}) = o\left(\frac{1}{t}\right)$

Second part $\bar{N}_k(t) = E(N_k(t)) = t p_k(t)$

Claim: $\delta > -n, m \geq 1; \exists C(m, \delta); \forall t \geq 1, \forall k \in \mathbb{N}$
 $|N_k(t) - p_k \cdot t| \leq C$

Exercise: Total degree of High Degree vertices

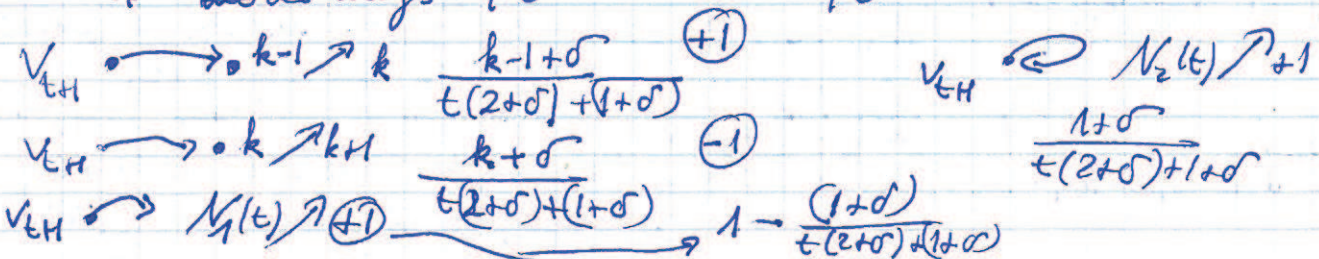
$l = l(t) \rightarrow \infty; t \rightarrow \infty, t \cdot l^{2-\gamma} \geq K \sqrt{t \log t}$ K large enough

$\exists B > 0$ s.t. $P(\forall i \text{ igen } l \text{-re } \sum_{D_i(t) \geq l} D_i(t) \geq B t l^{2-\gamma}) =$
 es $N_{\geq l}(t) \gg \sqrt{t} = 1 - o\left(\frac{1}{t}\right)$

Expected degree sequence $m=1$

$$E(N_k(t+1) | P_{A_t}) = N_k(t) + E(N_k(t+1) - N_k(t) | P_{A_t})$$

4 ~~addo~~ ways $\neq 0$



$$E(N_k(tH) - N_k(t) | \mathcal{P}_t) = \frac{k-1+\delta}{t(2+\delta)+(1+\delta)} \cdot N_{k-1}(t) - \frac{k+\delta}{t(2+\delta)+(1+\delta)} \cdot N_k(t)$$

$$k \geq 1, N_0(t) := 0 \quad + \mathbb{I}(k=1) \left(1 - \frac{1+\delta}{t(2+\delta)+(1+\delta)}\right) + \mathbb{I}(k=2) \frac{1+\delta}{t(2+\delta)+1+\delta}$$

⇒ taking expectation we get

$$\bar{N}_k(tH) = \bar{N}_k(t) + \frac{k-1+\delta}{t(2+\delta)+(1+\delta)} \bar{N}_{k-1}(t) - \frac{k+\delta}{t(2+\delta)+(1+\delta)} \bar{N}_k(t) + \mathbb{I}(k=1) \left(1 - \frac{1+\delta}{t(2+\delta)+(1+\delta)}\right) + \mathbb{I}(k=2) \frac{1+\delta}{t(2+\delta)+1+\delta}$$

let $t \rightarrow \infty$ $\left. \begin{array}{l} \bar{N}_k(tH) \sim (t+1)p_k \\ \bar{N}_k(t) \sim t \cdot p_k \end{array} \right\} \Rightarrow$ in the limit we get

II. megoldjuk:
$$p_k = \frac{k-1+\delta}{2+\delta} p_{k-1} - \frac{k+\delta}{(2+\delta)} p_k + \mathbb{I}(k=1) \quad (p_0=0)$$

$$p_k \left(\frac{2+\delta+k+\delta}{2+\delta} \right) = \frac{k-1+\delta}{2+\delta} p_{k-1} + \mathbb{I}(k=1)$$

$$p_k = \frac{k-1+\delta}{2+k+2\delta} p_{k-1} + \mathbb{I}(k=1) \frac{2+\delta}{k+2+2\delta} \quad (p_0=0)$$

$$p_1 = \frac{2+\delta}{1+2+2\delta} = \frac{2+\delta}{3+2\delta} ; k > 1 \quad p_k = \frac{k-1+\delta}{2+k+2\delta} p_{k-1}$$

$$\Rightarrow p_k = \prod_{j=2}^k \frac{j-1+\delta}{2+j+2\delta} \cdot \frac{2+\delta}{3+2\delta} = \frac{\Gamma(k+\delta)}{\Gamma(1+\delta)} \cdot \frac{\Gamma(4+2\delta)}{\Gamma(k+3+2\delta)} \cdot \frac{2+\delta}{3+2\delta} = \frac{\Gamma(k+\delta)}{\Gamma(k+3+2\delta)} \cdot \frac{\Gamma(3+2\delta)}{\Gamma(1+\delta)} \cdot (2+\delta) \quad \checkmark$$

Most össe alakulhat vissza a képlet. $E_k(t) := \bar{N}_k(t) - t p_k$

lém $\max_k |E_k(t)| \leq C$

2) $(t+1)p_k = t p_k + \frac{k-1+\delta}{(2+\delta)} p_{k-1} - \frac{k+\delta}{2+\delta} p_k + \mathbb{I}(k=1)$ [addig gyötörjél míg olyan nem lesz, mint I, és a hülöbörjé leír.]

$$= t p_k + \frac{k-1+\delta}{t(2+\delta)+(1+\delta)} p_{k-1} \cdot t - \frac{k+\delta}{t(2+\delta)+(1+\delta)} t p_k + \mathbb{I}(k=1) + \left(\frac{1}{2+\delta} - \frac{t}{t(2+\delta)+1+\delta} \right) (k-1+\delta) p_{k-1} - \left(\frac{1}{2+\delta} - \frac{t}{t(2+\delta)+1+\delta} \right) (k+\delta) p_k$$

$$f_k(t) = \frac{1+\sigma}{t(2+\sigma)+1+\sigma} \cdot (\mathbb{I}(k=2) - \mathbb{I}(k=1))$$

\$\Rightarrow\$ (*) - I yields

$$E_k(t+1) = \left(1 - \frac{k+\sigma}{t(2+\sigma)+1+\sigma}\right) E_k(t) + \frac{k-1+\sigma}{t(2+\sigma)+1+\sigma} E_{k-1}(t) + K_k(t) + f_k(t)$$

induction \$t\$-re. \$t=1, \bar{N}_k(1) = \mathbb{I}(k=2) \quad p_k \le 1\$

$$|E_k(1)| = |\bar{N}_k(1) - p_k| \leq \max\{\bar{N}_k(1), p_k\} \leq 1$$

\$k=1\$ &

$$|K_k(t)| \leq C_k \frac{1}{t+1} \quad f_k(t) \leq C_f \frac{1}{t+1} \quad C_f = 1, \quad C_k = \frac{(1+\sigma)(2+\sigma)}{2+\sigma}$$

$$E_0(t) = N_0(t) - p_0 = 0$$

\$\circ\$'s len.

$$|E_1(t+1)| = \left(1 - \frac{1+\sigma}{t(2+\sigma)+1+\sigma}\right) |E_1(t)| + |K_1(t)| + |f_1(t)|$$

$$\stackrel{\geq 0}{\leq} \leq C \cdot \frac{1+\sigma}{2+\sigma} - (C_k + C_f) \leq C_k \frac{1}{t+1} \leq C_f \frac{1}{t+1}$$

$$\frac{1}{t(2+\sigma)+1+\sigma} \leq \frac{1}{(t+1)(2+\sigma)}$$

$$E_1(t+1) \leq C - \frac{1}{t+1} \left(C \cdot \frac{1+\sigma}{2+\sigma} - (C_k + C_f) \right) \leq C \quad \text{he } C > \frac{2+\sigma}{1+\sigma} (C_k + C_f)$$

\$k=1 \checkmark\$ ke \$k=2\$.

\$k \ge 2\$-re

$$|E_k(t+1)| \leq \left(1 - \frac{k+\sigma}{t(2+\sigma)+1+\sigma}\right) |E_k(t)| + \frac{k-1+\sigma}{t(2+\sigma)+1+\sigma} |E_{k-1}(t)| + |K_k(t)| + |f_k(t)|$$

\$\circ\$'s len. \$k \le t(2+\sigma)+1\$

$$\leq C \left(1 - \frac{1}{t(2+\sigma)+1+\sigma}\right) + (C_k + C_f) \frac{1}{t+1} \leq$$

$$\leq C - \frac{1}{t+1} \left(\frac{C}{2+\sigma} - (C_k + C_f) \right) \leq C \quad C > (2+\sigma)(C_k + C_f)$$

\$k > t(2+\sigma)+1\$

max degree in \$PA_L(1, \sigma)\$ is \$(t+2) \Rightarrow \bar{N}_k(t+1) = 0 \quad k \ge t(2+\sigma)+2\$

$$\Rightarrow |E_k(t+1)| = (t+1)p_k$$

$$p_k \leq C_p \cdot \left(\frac{1}{t+1}\right)^{-(3+\sigma)} \leftarrow \phi(\sigma)$$

$$\Rightarrow |E_k(t+1)| = (t+1)p_k \leq C_p \cdot \left(\frac{1}{t+1}\right)^{-(2+\sigma)} \leq C_p$$

$$C := \max\left\{ (2+\sigma)(C_k + C_f), \frac{(2+\sigma)(C_k + C_f)}{1+\sigma}, C_p \right\}$$