

# WEEK 9

# Lecture 1

Last week we have introduced our first model of network growth

the BA model with PREFERENTIAL ATTACHMENT

$$\pi_i = \frac{k_i}{\sum_j k_j}$$

Visualisation in the case

$$m_0 = 2$$

$$m_0 = 1$$

$$m = 2$$

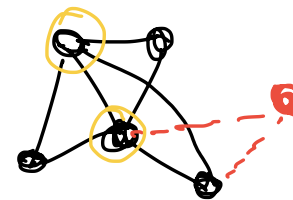
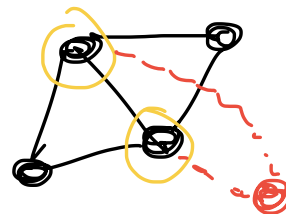
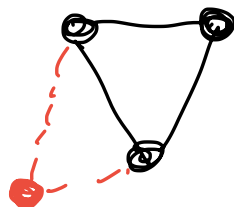
$$t = 1$$

$$t = 2$$

$$t = 3$$

$$t = 4$$

$$t = 5$$

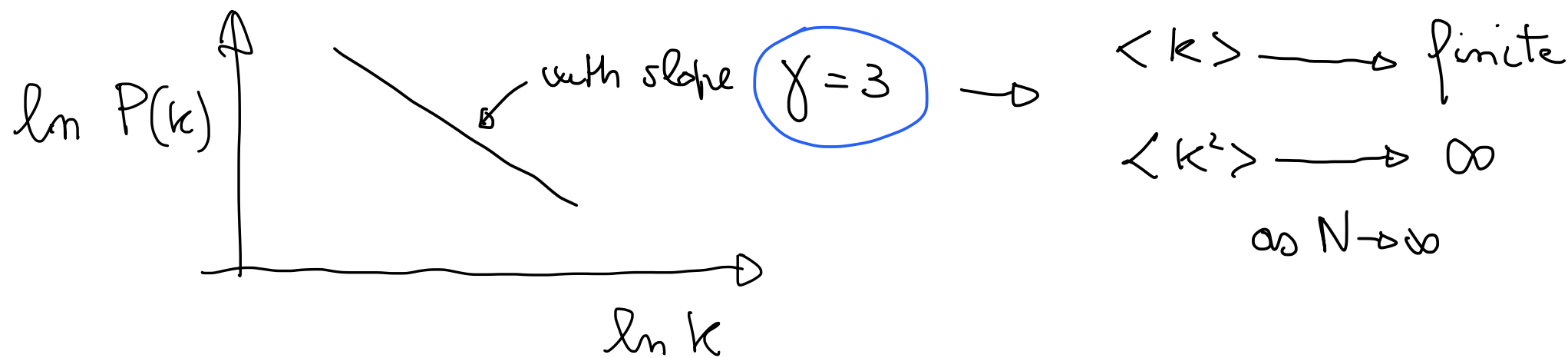


Last time we demonstrated that (in the MF approximation)

$$k_i(t) = m \left( \frac{t}{t_i} \right)^{\frac{1}{2}} \quad \text{for } t \geq t_i$$
$$t \gg 1$$



Today we are going to show the BA model is able to produce scale-free degree distributions



## PROPOSITION

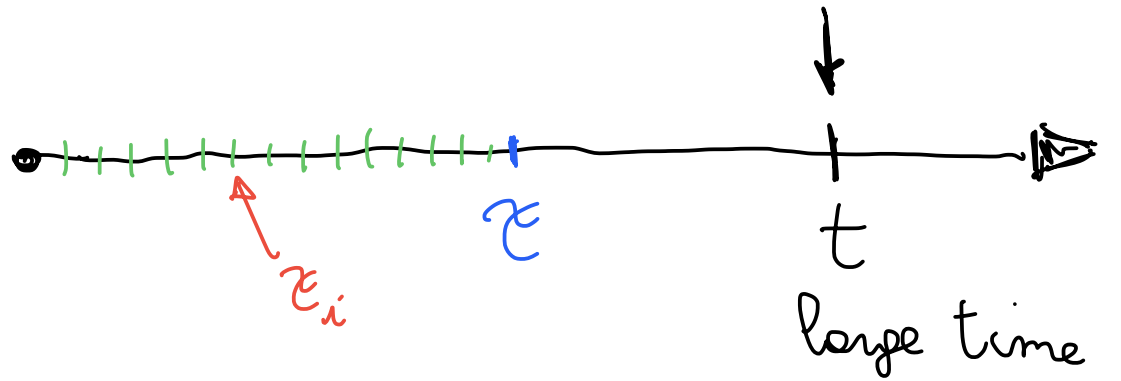
In the MF approximation the degree distribution of the BA model is given, in the limit  $t \rightarrow \infty$ , by:

$$P(k) \approx \frac{2m^2}{k^3} = \mathcal{O}(k^{-\gamma}) \text{ with } \gamma = 3$$



**Proof** At each time step a new node arrives in the network  
 Hence for  $t \gg 1$  the probability that a randomly chosen node has arrived at a time  $t_i < \tau$  can be approximated as:

$$\text{Prob}(t_i < \tau) \approx \frac{\tau}{t}$$



Let us write

$$\text{Prob}(k_i(t) > k) = \text{Prob}\left(m\left(\frac{t}{t_i}\right)^{\frac{1}{2}} > k\right) =$$

using  $\otimes \otimes$   $k_i(t) = m\left(\frac{t}{t_i}\right)^{\frac{1}{2}}$

$$= \text{Prob}\left(\sqrt{\frac{t}{t_i}} > \frac{k}{m}\right) = \text{Prob}\left(\frac{t}{t_i} > \frac{k^2}{m^2}\right) = \text{Prob}\left(t_i < t \frac{m^2}{k^2}\right)$$

Hence

$$\text{Prob}(k_i(t) > k) = \text{Prob}\left(t_i < \underbrace{t \frac{m^2}{k^2}}_{\tau}\right) \approx \frac{t \frac{m^2}{k^2}}{t} = \frac{m^2}{k^2}$$

$$\text{Prob}(k_i(t) > k) \approx \frac{m^2}{k^2}$$

This is the probability that, if we observe the network a time  $t$ , a randomly chosen mode has a degree larger than the value  $k$

The degree distribution  $P(k)$ , i.e. the probability that a randomly chosen node  $i$  has degree  $k$ , can be written as

$$P(k) = \frac{d}{dk} \text{Prob}(k_i(t) \leq k) =$$

$$= \frac{d}{dk} \left[ 1 - \text{Prob}(k_i(t) > k) \right] =$$

$$= \frac{d}{dk} \left[ 1 - \frac{m^2}{k^2} \right] = -m^2 \frac{d}{dk} k^{-2} = \frac{2m^2}{k^3}$$

$$P(k) \approx \frac{2m^2}{k^3} = C k^{-\gamma} \text{ with } \gamma = 3$$

which is the formula  
~~we~~ ~~wanted~~ ~~to~~ find

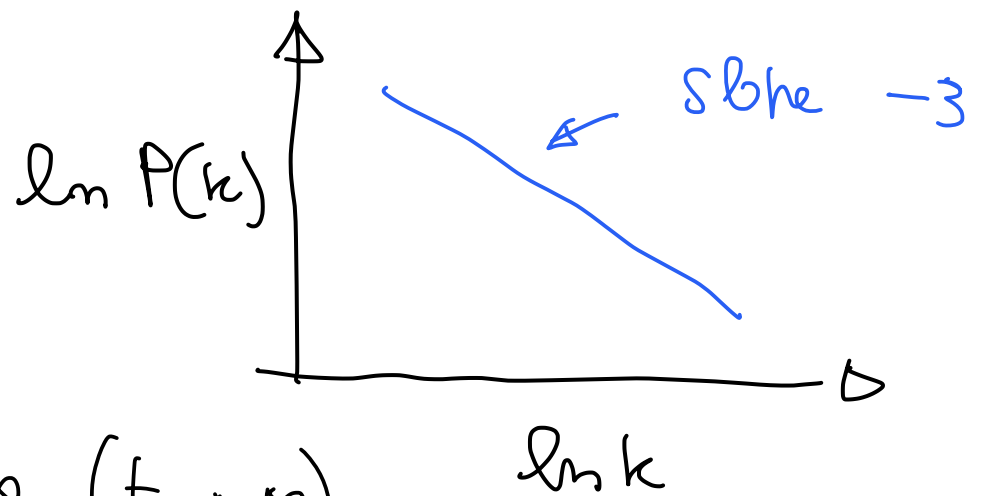
BA model produces SCALE-FREE

networks with  $\gamma$  in the range

$(2, 3]$  where we have

$\langle k \rangle \rightarrow$  finite

$\langle k^2 \rangle \rightarrow \infty$  when  $N \rightarrow \infty$  ( $t \rightarrow \infty$ )



The M.F. is a strong approximation.

I now present a more rigorous approach that provides exact results for the degree distribution in the limit  $t \rightarrow \infty$

## PROPOSITION | MASTER EQUATION

The master equation is the equation that describes the evolution of the average number  $N_k(t)$  of nodes with degree  $k$  at time  $t$ .

In the case of the BA model the master equation reads

$$N_k(t+1) = N_k(t) + \tilde{\Pi}(k-1) N_{k-1}(t) - \tilde{\Pi}(k) N_k(t)$$

for  $k > m$

$$N_k(t+1) = N_k(t) + 1 - \tilde{\Pi}(k) N_k(t)$$

for  $k = m$

where  $\tilde{\Pi}(k)$  is the probability that a node of degree  $k$  acquires a link at time  $t$

$$\tilde{\Pi}(k) = m \frac{k}{\sum_{k'} k' \cdot N_{k'}(t)}$$

Proof:

We first observe that

$$\tilde{\Pi}(k) = m \prod_i \Big|_{k_i=k} = m \frac{k}{\sum_j k_j} = m \frac{k}{\sum_{k'} k' N_{k'}(t)}$$

probability of linking to a node  $i$  of degree  $k_i=k$

$$\sum_{k'} k' \cdot N_{k'}(t)$$

To derive  $\otimes \otimes$  we consider separately the cases  $\begin{cases} k > m \\ k = m \end{cases}$

$k > m$

$$N_k(t+1) = N_k(t) + \text{GAIN TERM} - \text{LOSS TERM}$$

GAIN TERM  
 $k-1 \rightarrow k$

$$\tilde{\Pi}(k-1) \cdot N_{k-1}(t)$$

the average # of nodes of degree  $k-1$  at time  $t$  that acquire a link at time  $t+1$

LOSS TERM  
 $k \rightarrow k+1$

$$\tilde{\Pi}(k) \cdot N_k(t)$$

average # of nodes of degree  $k$  at time  $t$  that acquire a link at time  $t+1$

$$N_k(t+1) = N_k(t) + \tilde{\Pi}(k-1) N_{k-1}(t) - \tilde{\Pi}(k) N_k(t)$$

$k > m$

$k = m$

$$N_m(t+1) = N_m(t) + \text{GAIN TERM} - \text{LOSS TERM}$$

GAIN TERM  
 addition of  
 a new node  
 of degree  $k = m$

1

At each time step one new node of degree  $m$  is added to the network

LOSS TERM  
 $m \rightarrow m+1$

$$\tilde{\Pi}(m) \cdot N_m(t)$$

The average # of nodes of degree  $m$  at time  $t$  that acquire a new link

$$\underline{N_m(t+1) = N_m(t) + 1 - \tilde{\Pi}(m) \cdot N_m(t)}$$

or

$$\underline{N_k(t+1) = N_k(t) + 1 - \tilde{\Pi}(k) \cdot N_k(t)}$$

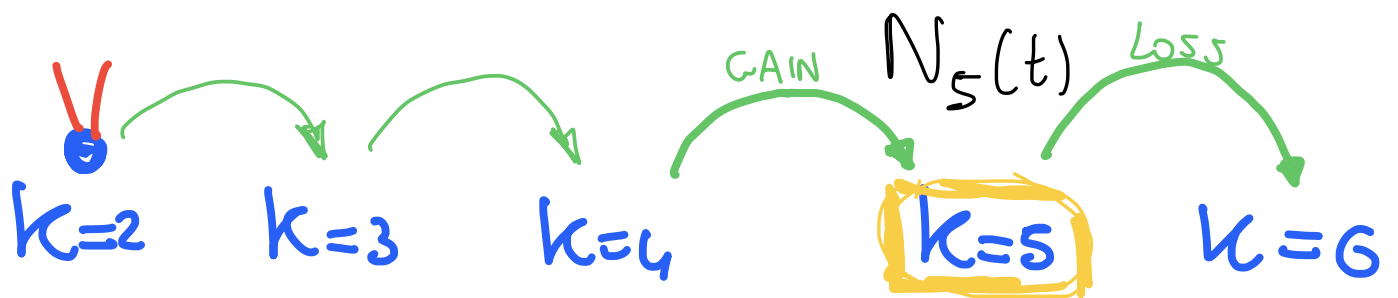
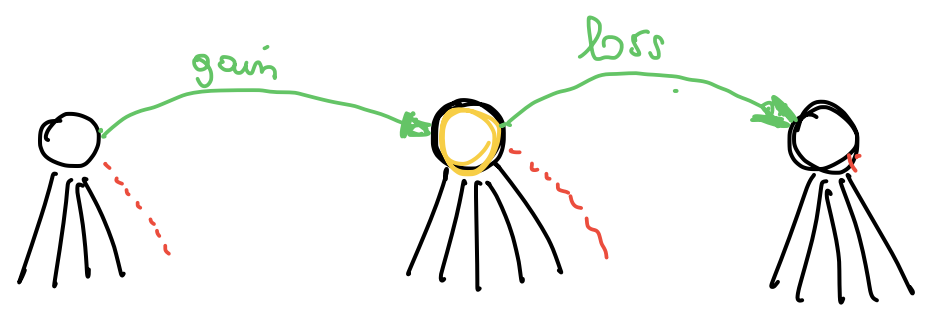
$k=m$

The equations for  $k > m$  and  $k = m$  can be written in a compact form

$$N_k(t+1) = N_k(t) + \tilde{\Pi}(k-1) N_{k-1}(t) \left[ \underset{\substack{1-1 \\ \uparrow \text{if } k=m}}{1 - \delta_{k,m}} \right] - \tilde{\Pi}(k) N_k(t) + \underset{\uparrow 1}{\delta_{k,m}}$$

where  $\delta_{k,m} = \begin{cases} 1 & \text{if } k=m \\ 0 & \text{if } k \neq m \end{cases}$  for  $k \geq m$

KRONECKER  
DELTA



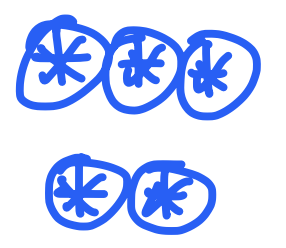


# PROPOSITION $P(k)$

( $N \rightarrow \infty$ )

The degree distribution  $P(k)$  of the BA model, in the limit  $t \rightarrow \infty$  is given by:

$$P(k) = \frac{2m(m+1)}{k(k+1)(k+2)} \quad \text{for } k \geq m$$



Therefore for  $k \gg 1$   $P(k) \sim k^{-3}$  with  $\gamma = 3$

Proof

$$\tilde{\Pi}(k) = m \frac{k}{\sum_j k_j} = m \frac{k}{\sum_{k'} k' N_{k'}(t)} \approx \frac{mk}{2mt} = \frac{k}{2t}$$

For  $t \gg 1$   $\sum_j k_j \approx 2mt$

Hence for  $t \gg 1$  the master equation reads:

$$\begin{cases} N_k(t+1) = N_k(t) + \frac{k-1}{2t} N_{k-1}(t) - \frac{k}{2t} N_k(t) & k > m \\ N_m(t+1) = N_m(t) + 1 - \frac{m}{2t} N_m(t) & k = m \end{cases}$$

Assuming that for large times ( $t \gg 1$ ) the degree distribution converges to a distribution  $P(k)$

For  $t \gg 1$   $N_k(t) \approx N P(k) \approx t P(k)$

$N_k(t) = t P(k)$

at  $t \gg 1$

for  $t \gg 1$   $N \approx t$

Hence we can rewrite the master equation in terms of  $P(k)$

$$\begin{cases} (t+1) P(k) = t P(k) + \frac{k-1}{2t} t P(k-1) - \frac{k}{2t} t P(k) & k > m \\ (t+1) P(m) = t P(m) + 1 - \frac{m}{2t} t P(m) & k = m \end{cases}$$

$$\begin{cases} \left(1 + \frac{k}{2}\right) P(k) = \frac{k-1}{2} P(k-1) \\ \left(1 + \frac{m}{2}\right) P(m) = 1 \end{cases}$$

$$\begin{aligned} P(k) &= \frac{k-1}{k+2} P(k-1) \\ P(m) &= \frac{1}{1 + \frac{m}{2}} = \frac{2}{m+2} \end{aligned}$$

$$\begin{cases} P(k) = \frac{k-1}{k+2} P(k-1) & k > m \\ P(m) = \frac{2}{m+2} \end{cases}$$

$$P(k) = \frac{k-1}{k+2} P(k-1) = \frac{k-1}{k+2} \frac{k-2}{k+1} P(k-2) = \frac{k-1}{k+2} \frac{k-2}{k+1} \frac{k-3}{k} P(k-3) = \dots =$$

$$= \frac{\cancel{k-1}}{\textcircled{k+2}} \frac{\cancel{k-2}}{\textcircled{k+1}} \frac{\cancel{k-3}}{\textcircled{k}} \frac{\cancel{k-4}}{\cancel{k-1}} \dots \frac{\cancel{m+3}}{\cancel{m+6}} \frac{\textcircled{m+2}}{\cancel{m+5}} \frac{\textcircled{m+1}}{\cancel{m+4}} \frac{\textcircled{m}}{\cancel{m+3}} P(m)$$

From  $P(k) = \frac{k-1}{k+2} P(k-1)$

with  $k-1 = m$   
 $k = m+1$   
 $k+2 = m+3$

$$P(m+1) = \frac{m}{m+3} P(m)$$

with  $k-1 = m+1$   
 $k = m+2$   
 $k+2 = m+4$

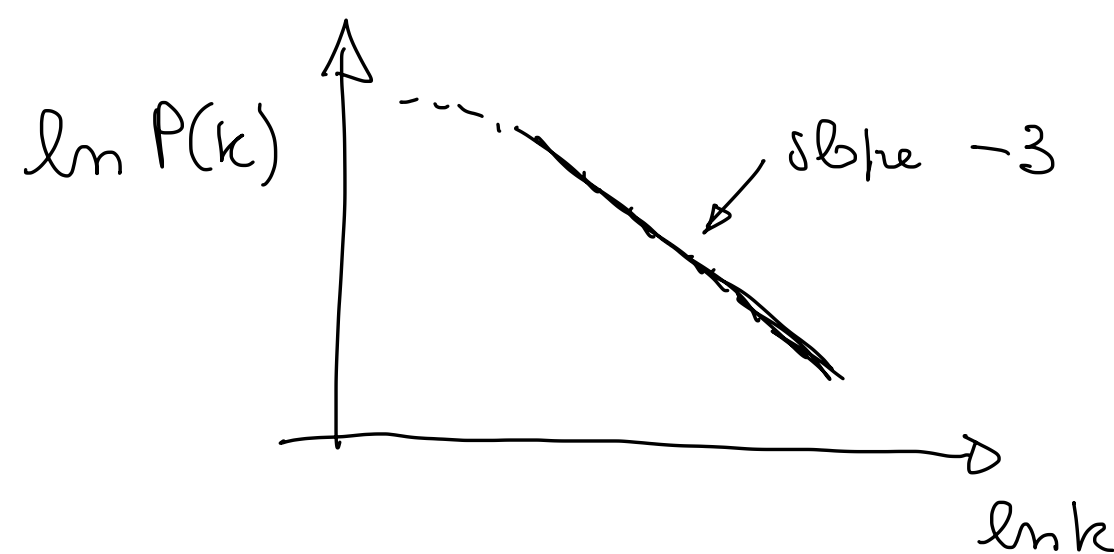
$$P(m+2) = \frac{m+1}{m+4} P(m+1)$$

$$P(m+2) = \frac{m+1}{m+4} \frac{m}{m+3} P(m)$$

$$P(k) = \frac{(m+2)(m+1)m}{(k+2)(k+1)k} \cdot \textcircled{P(m)} = \frac{\cancel{(m+2)}(m+1)m}{(k+2)(k+1)k} \frac{2}{\cancel{m+2}}$$

$\frac{2}{m+2}$

$$P(k) = \frac{2m(m+1)}{k(k+1)(k+2)} \quad \text{for } k \geq m$$



Remember that in the MF approximation we had

$$P(k) \approx \frac{2m^2}{k^3} \quad (\text{not very different !!})$$

IMPORTANT : Submit your course work (PDF)

Assessed course work 4

closes on Wed 20 March

at 5 PM