

WEEK 5 Lecture 1

CHAPTER 4 RANDOM GRAPHS

4.1 INTRODUCTION

Real-world
networks

vs

RANDOM GRAPHS with same N, L
and disordered arrangement of
links

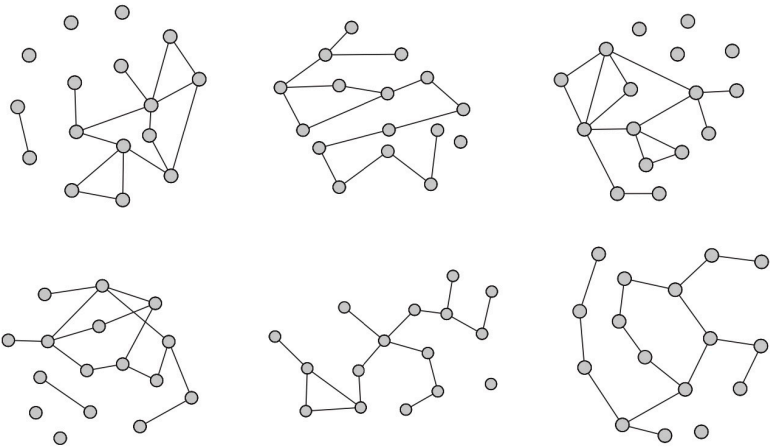
2 pioneering papers by ERDŐS and RENYI in 1959
1960

"On the evolution of random graphs"

PROBABILITY + GRAPH THEORY

IDEA: To study the properties of graphs a function of the increasing # of random connections

ENSEMBLE of GRAPHS



Six different realisations of model A with $N = 16$ and $K = 15$.

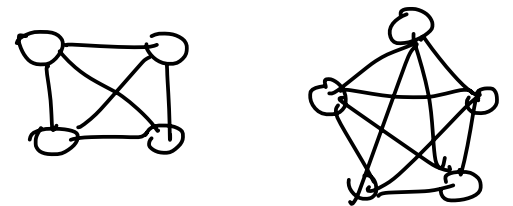
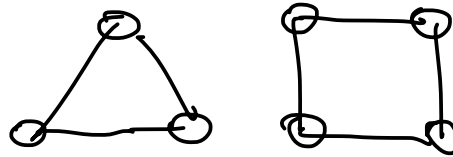
$$N = 16 \quad L = 15$$

→ the average size of the LARGEST COMPONENT

→ the average # of subgraphs:

CYCLES

CLIQUEES



4.2 RANDOM GRAPH ENSEMBLES

$\mathbb{G}(N, L)$ ensemble considers every simple network with N nodes and L links with equal probability

DEF

$\mathbb{G}(N, L)$ ← fix L

ENSEMBLE A

The $G(N, L)$ ensemble assigns to each simple network

$G = (V, E)$ a probability

$$P(G) = \begin{cases} \frac{1}{Z} & \text{if } |V| = N \text{ and } |E| = L \\ 0 & \text{otherwise} \end{cases}$$

$Z = \#$ of simple networks with N nodes and L links

- The max # of links M in a simple network is

$$M = \binom{N}{2} = \frac{N(N-1)}{2}$$

binomial coefficient $\binom{m}{k} = \frac{m!}{k!(m-k)!}$

- Hence Z is given by the # of ways in which we can choose L links out of M possibilities:

$$Z = \binom{M}{L} = \binom{\frac{N(N-1)}{2}}{L}$$

ENSEMBLE B

$\mathbb{G}(N, p)$ ensemble considers every simple network with N nodes obtained by connecting each pair of nodes with a probability p

DEF $\mathbb{G}(N, p)$ ← fix $p : 0 \leq p \leq 1$

The $\mathbb{G}(N, p)$ ensemble assigns to each simple network

$G = (V, E)$ with $|V| = N$ nodes a probability

$$P(G) = p^L (1-p)^{\frac{N(N-1)}{2} - L} \quad \text{where } L = |E|$$

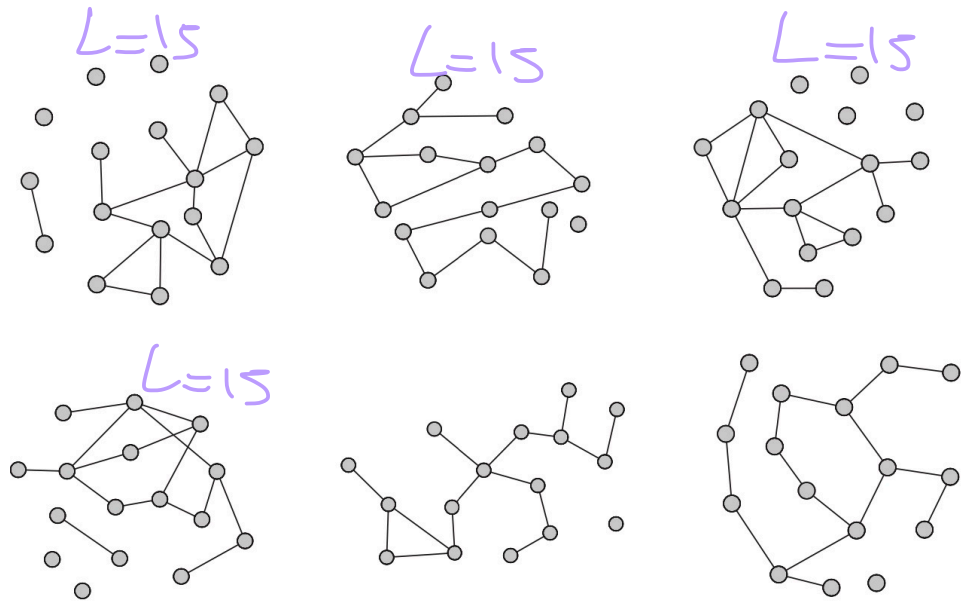
Each network in $\mathbb{G}(N, p)$ has N fixed but L can take different values, and it can be seen as the result of $M = \frac{N(N-1)}{2}$

independent coin tossings, one for each pair of nodes, with success probability (of drawing a link)

probability that L pairs of nodes are linked

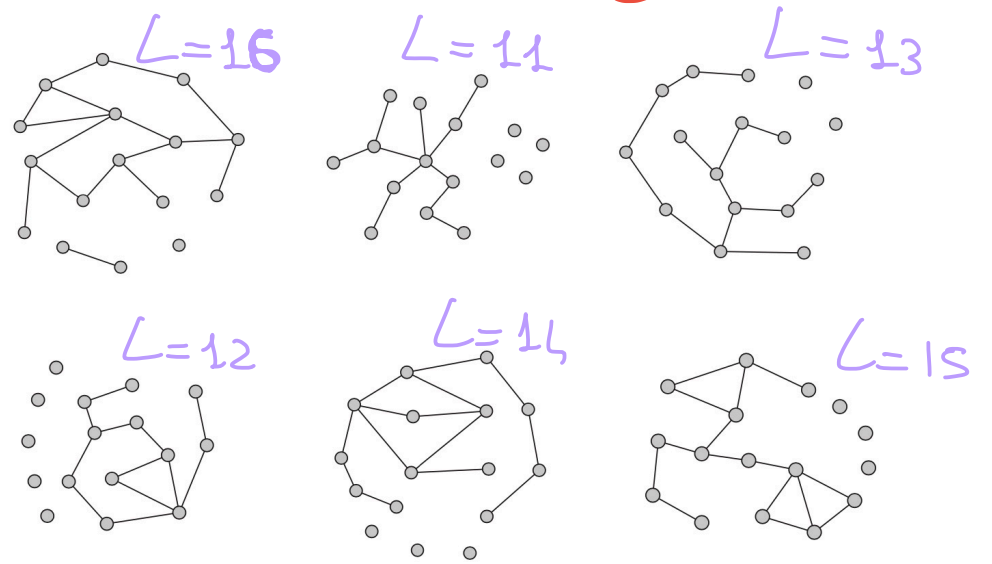
probability that the remaining $M - L$ pairs are not linked

$\mathbb{G}(N, L)$
ENSEMBLE A



Six different realisations of model A with $N = 16$ and $K = 15$.

$\mathbb{G}(N, p)$
ENSEMBLE B



Six different realisations of model B, with $N = 16$ and $p = 0.125$.

$N = 16$ $L = 15$

$N = 16$ $p = 0.125$
 $p \cdot \frac{N(N-1)}{2} = 15$

$\mathbb{G}(N, L)$ and $\mathbb{G}(N, p)$ are ASYMPTOTICALLY EQUIVALENT

For $L = p \frac{N(N-1)}{2}$ and $N \rightarrow \infty$ then the 2 ensembles

$\mathbb{G}(N, L)$ and $\mathbb{G}(N, p)$ have the same statistical properties

Hence we focus on $\mathbb{G}(N, p)$

4.3 DISTRIBUTION P_L

PROPOSITION

The probability P_L that a network in $\mathbb{G}(N, p)$ has L links follows the BINOMIAL DISTRIBUTION

$$P_L = \binom{\frac{N(N-1)}{2}}{L} p^L (1-p)^{\frac{N(N-1)}{2} - L}$$

i.e. $L \sim \text{Bin} \left(\frac{N(N-1)}{2}, p \right)$

BINOMIAL DISTRIBUTION

Proof

probability that L pairs of nodes are linked

probability that the remaining pairs are NOT linked

of ways we can choose the L linked pairs of nodes out of the M possibilities

Since $L \sim \text{Bin} \left(\frac{N(N-1)}{2}, p \right)$

$$\langle L \rangle = \frac{N(N-1)}{2} \cdot p$$

↑
average # of links
in a network of the ensemble $G(N, p)$

4.4 DEGREE DISTRIBUTION!

PROPOSITION

The degree distribution $P(k)$ of the $G(N, p)$ ensemble is binomial

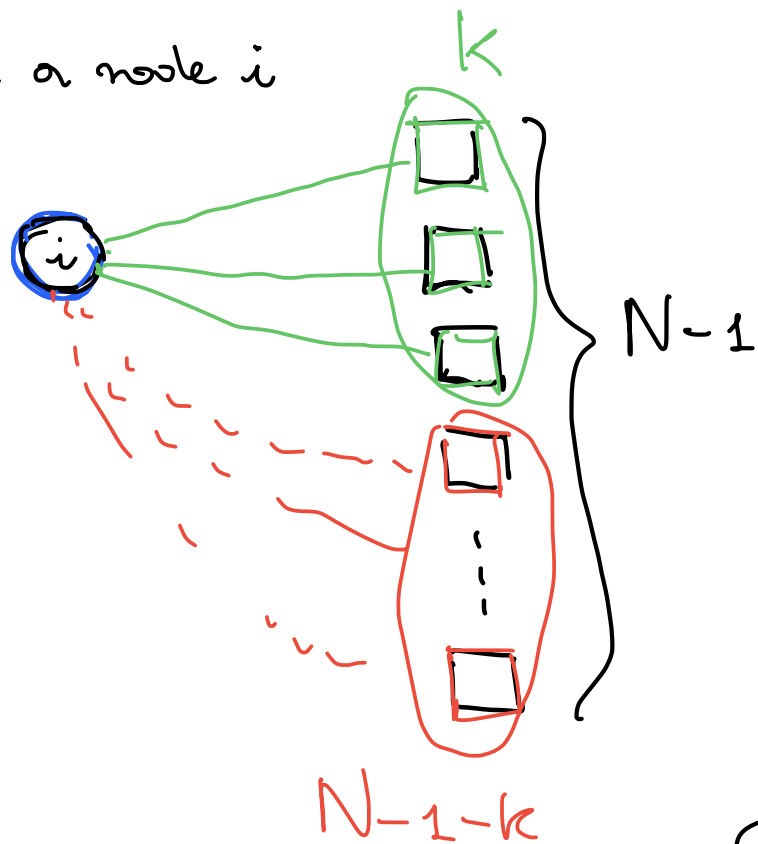
$$P_B(k) = \binom{N-1}{k} p^k (1-p)^{N-1-k}$$

i.e. $k \sim \text{Bin}(N-1, p)$

$$\binom{N-1}{k}$$

↑
probability of linking
two nodes

Proof: Take a node i



P^k is the probability of the existence of k links

$(1-P)^{N-1-k}$ is the probability for the absence of the remaining links

$\binom{N-1}{k} = \#$ of different ways of selecting the end points of the k links among the $N-1$ possibilities

GENERATING FUNCTIONS

DEF

Given a degree distribution $P(k)$, its GENERATING FUNCTION

$G(x)$ is defined by

$$G(x) = \sum_{k=0}^{\infty} P(k) x^k$$

PROPOSITION

The generating function $G(x)$ of $P(k)$ has the following properties:

① $G(1) = 1$

② $G^{[m]}(1) \equiv \left. \frac{d^m G(x)}{dx^m} \right|_{x=1} = \langle k(k-1)\dots(k-m+1) \rangle$

m -th FACTORIAL
MOMENT of $P(k)$

$\langle k \rangle = G'(1)$

$\langle k(k-1) \rangle = G''(1) \equiv G^{[2]}(1)$

$$G(x) = \sum_{k=0}^{\infty} P(k) x^k$$

Proof

① $G(1) = \sum_{k=0}^{\infty} P(k) 1^k = \sum_{k=0}^{\infty} P(k) = 1$

② $m=1$ $\frac{dG(x)}{dx} = \frac{d}{dx} \sum_{k=0}^{\infty} P(k) x^k = \sum_{k=0}^{\infty} P(k) \frac{d}{dx} x^k =$

$$= \sum_{k=0}^{\infty} P(k) k x^{k-1}$$

$$\left. \frac{dG(x)}{dx} \right|_{x=1} = \sum_{k=0}^{\infty} P(k) k = \langle k \rangle$$

$m=2$

$$\frac{d^2 G(x)}{dx^2} = \sum_{k=0}^{\infty} P(k) \frac{d^2}{dx^2} x^k = \sum_{k=0}^{\infty} P(k) k(k-1) x^{k-2}$$

$$\left. \frac{d^2 G(x)}{dx^2} \right|_{x=1} = \sum_{k=0}^{\infty} P(k) k(k-1) = \langle k(k-1) \rangle$$

and so on

BINOMIAL DISTRIBUTION

$$P(k) = \text{Bin}(N-1, p) = \binom{N-1}{k} p^k (1-p)^{N-1-k}$$

$k=0, 1, \dots, N-1$

$$G_B(x) = \sum_{k=0}^{\infty} P(k) x^k$$

10

$$= \sum_{k=0}^{N-1} \binom{N-1}{k} p^k x^k (1-p)^{N-1-k}$$

Recall the NEWTON BINOMIAL FORMULA

$$(a+b)^M = \sum_{k=0}^M \binom{M}{k} a^k b^{M-k}$$

We use it with $a = px$ $b = 1-p$ $M = N-1$

$$G_B(x) = (px + 1-p)^{N-1}$$

GENERATING FUNCTION of Binomial $P(k)$

We can now use property (2) of generating functions to calculate the moments of $P(k)$

$$\langle k \rangle = G_B'(1) = (N-1) (px + 1-p)^{N-2} \cdot p \Big|_{x=1} = (N-1) (p + 1-p)^{N-2} \cdot p$$

$$\langle k \rangle = (N-1)p$$

$$\langle k(k-1) \rangle = G_B^{[2]}(1) = (N-1)(N-2) (px + 1-p)^{N-3} \cdot p^2 \Big|_{x=1} =$$

(11)

$$= (N-1)(N-2) p^2$$

$$\langle k(k-1) \rangle = (N-1)(N-2) p^2$$

$$\langle k^2 - k \rangle = \langle k^2 \rangle - \langle k \rangle$$

$$\langle k \rangle = (N-1)p$$

$$\langle k^2 \rangle = \langle k \rangle + \langle k^2 - k \rangle = \langle k \rangle + (N-1)(N-2) p^2$$

$$= (N-1)p + (N-1)(N-2) p^2 =$$

$$= (N-1)p [1 + (N-2)p]$$

VARIANCE $\sigma^2 \equiv \langle k^2 \rangle - \langle k \rangle^2 = (N-1)p [1 + (N-2)p - (N-1)p]$

$$= (N-1)p [1 - 2p + p] = (N-1)p(1-p)$$

$$\sigma^2 = (N-1)p(1-p)$$

Analogously

$$\langle k(k-1)\dots(k-m+1) \rangle = \frac{(N-1)!}{(N-m-1)!} p^m$$

POISSON DISTRIBUTION

$$P(k) = \text{Poiss}(e) = \frac{c^k e^{-c}}{k!} \quad k=0,1,\dots$$

$$G_P(x) = \sum_{k=0}^{\infty} P(k) x^k = e^{-c} \sum_{k=0}^{\infty} \frac{x^k c^k}{k!}$$

RECALL the TAYLOR series of the exponential

$$e^{\alpha} = \sum_{k=0}^{\infty} \frac{\alpha^k}{k!}$$

We use it with $\alpha = x \cdot c$

We get:

$$G_P(x) = e^{-c} \cdot e^{cx}$$

GENERATING
FUNCTION
of POISSON $P(k)$

We can calculate the MOMENTS of Poiss(c)

$$\langle k \rangle = G_P'(1) = e^{-c} e^{cx} \Big|_{x=1} = e^{-c} e e^c = c$$

$$\langle k \rangle = c$$

$$\langle k(k-1) \rangle = G_p''(1) = \left. e^{-c} c^2 e^{cx} \right|_{x=1} = c^2$$

$$\langle k(k-1) \rangle = c^2$$

Analogously

$$\langle k(k-1) \dots (k-m+1) \rangle = c^m$$

IMPORTANT: Please do not forget to submit
Quiz 2