

RANDOM

GRAPHS

Basic Models

$$G_{n,m} = ([n], E_{n,m})$$

Vertex set $[n] = \{1, 2, \dots, n\}$

Each graph with m edges

has same probability

$$\frac{1}{\binom{N}{m}},$$

$$N = \binom{n}{2}$$

$$G_{n,p} = ([n], E_{n,p})$$

Vertex set $[n]$

$$P_r(G_{n,p} = G) = p^{|E(G)|} (1-p)^{N - |E(G)|}$$

i.e. each edge occurs independently
with probability p .

Graph property \mathcal{P} .

$$p = \frac{m}{N}$$

$$P_r(G_{n,p} \in \mathcal{P}) = \sum_{\mu=0}^N P_r(G_{n,p} \in \mathcal{P} \mid |E_{n,p}| = \mu) \times P_r(|E_{n,p}| = \mu)$$

$$= \sum_{\mu=0}^N P_r(G_{n,\mu} \in \mathcal{P}) P_r(|E_{n,p}| = \mu)$$

$$\geq P_r(G_{n,m} \in \mathcal{P}) P_r(|E_{n,p}| = m)$$

$$P(|E_{n,p}| = m) = \binom{N}{m} p^m (1-p)^{N-m}$$

$$\approx (1+o(1)) \frac{N^N \sqrt{2\pi N} p^m (1-p)^{N-m}}{m^m (N-m)^{N-m} 2\pi \sqrt{m(N-m)}}$$

$m \rightarrow \infty$
 $N-m \rightarrow \infty$

$$\approx (1+o(1)) \sqrt{\frac{N}{2\pi m(N-m)}}$$

$$\geq \frac{1}{10\sqrt{m}}$$

So,

$$\Pr(G_{n,m} \in \mathcal{P}) \leq 10m^{1/2} \Pr(G_{n,p} \in \mathcal{P}).$$

Monotone Properties

A property is monotone increasing \nearrow if

$$G \in \mathcal{P} \implies G + e \in \mathcal{P}$$

e.g. connectivity

Monotone decreasing \searrow if

$$G \in \mathcal{P} \implies G - e \in \mathcal{P}$$

e.g. planarity

Suppose \mathcal{P} is \nearrow . $\rho = \frac{1}{\sqrt{3}}$

$$\begin{aligned} P_r(G_{n,\rho} \in \mathcal{P}) &= \sum_{\mu=0}^N P_r(G_{n,\mu} \in \mathcal{P}) P_r(|E_{n,\rho}| = \mu) \\ &\geq P_r(G_{n,m} \in \mathcal{P}) \underbrace{\sum_{\mu=m}^N P_r(|E_{n,\rho}| = \mu)}_{\geq \frac{1}{2} - o(1)}. \end{aligned}$$

Central Limit Theorem \Rightarrow $\downarrow \geq \frac{1}{2} - o(1)$

$$P_r(G_{n,m} \in \mathcal{P}) \leq 3 P_r(G_{n,\rho} \in \mathcal{P}).$$

Graph Process:

$$G_0 = ([n], \emptyset), G_1, G_2, \dots, G_m, \dots, G_N = K_n$$

$$G_{m+1} = G_m \text{ plus random edge}$$

G_m and $G_{n,m}$ have same
distribution.

Markov Inequality

$X \geq 0$ is a random variable with finite mean μ .

$$P(X \geq t) \leq \frac{\mu}{t}$$

Proof

$$\begin{aligned} E(X) &= E(X|X < t) P(X < t) \\ &\quad + E(X|X \geq t) P(X \geq t) \\ &\geq t P(X \geq t). \end{aligned}$$



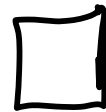
Chebyshev Inequality

X is a random variable with finite mean μ and variance σ^2 .

$$P(|X - \mu| \geq t) \leq \frac{\sigma^2}{t^2}$$

Proof

$$\begin{aligned} P(|X - \mu| \geq t\sigma) &= P((X - \mu)^2 \geq t^2\sigma^2) \\ &\leq \frac{E((X - \mu)^2)}{t^2\sigma^2} \\ &= \frac{\sigma^2}{t^2}. \end{aligned}$$



First Moment Method

Let X be a random variable with finite mean taking values in $\{0, 1, 2, \dots\}$.

$$P_r(X \neq 0) \leq E(X)$$

Proof

$$\begin{aligned} P_r(X \neq 0) &= P_r(X \geq 1) \\ &\leq \frac{E(X)}{1}. \end{aligned}$$

□

Second Moment Method

Let X be a non-negative random variable with finite mean and variance. Then

$$P(X > 0) \geq \frac{E(X)^2}{E(X^2)}$$

Proof

$$\text{Let } Y = \begin{cases} 0 & \text{if } X = 0 \\ 1 & \text{if } X > 0 \end{cases}$$

$$\text{So } XY = X$$

Cauchy-Schwarz inequality* implies

$$E(XY)^2 \leq E(X^2) E(Y^2)$$

or

$$E(X)^2 \leq E(X^2) P(X > 0).$$



* Consider quadratic $E((X + tY)^2) \geq 0$, as a function of t .

Evolution of a random graph.

We look at how $G_0, G_1, \dots, G_m, \dots$
evolves.

$\omega = \omega(n)$ denotes some slowly growing function e.g. $\omega = \log n$.

$$(1) \quad m \leq n^{1/2} / \omega$$

G_m is a matching whp

whp: with high probability
i.e. with probability $1 - o(1)$
as $n \rightarrow \infty$.

Let $p = \frac{m}{N}$ and let $X_2 =$ number of paths of length 2 in $G_{n,p}$.

$$\begin{aligned} E_p(X_2) &= 3 \binom{n}{3} p^2 \\ &\leq \frac{n^3 \times n}{N^2 \times w^2} \end{aligned}$$

$\rightarrow 0$.

$P_1(G_{n,p} \text{ contains path of length 2}) = o(1)$
monotone property

\Downarrow
 $P_1(G_{n,m} \text{ contains a path of length 2}) = o(1)$.

$$(ii) \quad m = \omega n^{1/2}, \quad m = o(n).$$

G_m contains a path of length 2 whp

Let $p = \frac{m}{n}$ and $X_2 = \#$ paths length 2.

$$\mathbb{E}(X_2) = 3 \binom{n}{3} p^2$$

$$\approx 2\omega^2$$

$$\rightarrow \infty$$

Does not imply $X_2 \neq 0$ whp.

Let \mathcal{P}_2 be the set of all paths of length two in K_n .

Let $\hat{X}_2 = \#$ of **isolated** paths of length 2 in $G_{n,p}$

$$\hat{X}_2 = \sum_{P \in \mathcal{P}_2} \mathbb{1}_{P \subseteq G_{n,p}}$$

$$E(\hat{X}_2) = 3 \binom{n}{3} p^2 (1-p)^{3(n-3)}$$

$$\geq (1-o(1)) \frac{n^3}{2} \cdot \frac{4p^2 n}{n^4} \cdot (1-6np)$$

$$\rightarrow \infty.$$

$n = o(n)$
 \Downarrow
 $np = o(1)$

$$\hat{X}_2^2 = \sum_{P \in \mathcal{P}_2} \sum_{Q \in \mathcal{P}_2} \mathbb{1}_{P \in G_{n,p}} \mathbb{1}_{Q \in G_{n,p}}$$

$$= \sum_{P, Q \in \mathcal{P}_2} \mathbb{1}_{P \in G_{n,p}} \mathbb{1}_{Q \in G_{n,p}}$$

$P=Q$ or P, Q vertex disjoint

$$E(\hat{X}_2^2) = \sum_P \left\{ \sum_Q P_r(G_{n,p} \supseteq Q \mid G_{n,p} \supseteq P) \right\} \times P_r(G_{n,p} \supseteq P)$$

Expression inside $\{\}$ is same for all P .

$$= E(\hat{X}_2) \left(1 + \sum_{\substack{Q \cap \{1,2,3\} \\ = \emptyset}} P_r(G_{n,p} \supseteq Q \mid G_{n,p} \supseteq \checkmark_2) \right)$$

$$\leq E(\hat{X}_2) \left(1 + \binom{n}{3} p^2 (1-p)^{3(n-6)} \right)$$

$$\leq E(\hat{X}_2) \left(1 + (1-p)^{-3} E(\hat{X}_2) \right)$$

* Conditioning means no edge to $\{1,2,3\}$

So

$$P_r(\hat{X}_2 \neq 0) \geq \frac{E(\hat{X}_2)^2}{E(\hat{X}_2)(1 + (1-p)^{-3} E(\hat{X}_2))}$$

$$= \frac{1}{(1-p)^{-3} + E(\hat{X}_2)^{-1}}$$

$$\rightarrow 1.$$

not monotone

Thus $P_r(G_{n,p} \geq \text{isolated 2-path}) \rightarrow \underline{1}$

$P_r(G_{n,p} \geq 2\text{-path}) \rightarrow \underline{1}$

$P_r(G_m \geq 2\text{-path}) \rightarrow \underline{1}$

monotone

$$P_1(G_m \ni 2\text{-path}) = \begin{cases} 0(1) & m \ll n^{1/2} \\ 1 - 0(1) & m \gg n^{1/2} \end{cases}$$

We say that $n^{1/2}$ is the

threshold

for the existence of a 2-path in $G_{n,m}$

Probability "jumps" from
 ~ 0 to ~ 1