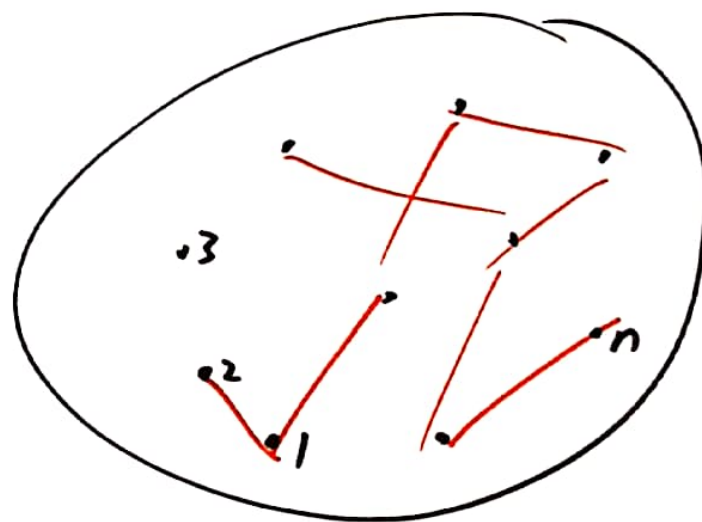


Definition For $n \geq 1$ an integer and $0 \leq p \leq 1$, the random graph $G(n, p)$ is the random graph on $[n] = \{1, 2, \dots, n\}$ in which each of the $\binom{n}{2}$ possible edges is present indep. with prob. p .



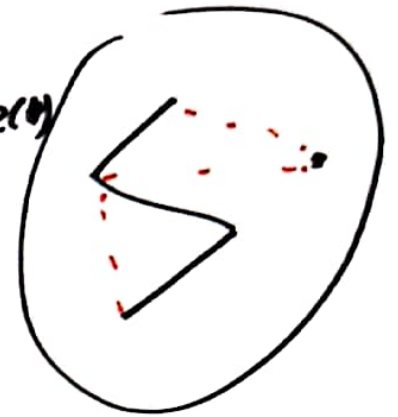
$$|V| = n$$

$$V = [n]$$

Or $\Omega = \{ \text{all } 2^{\binom{n}{2}} \text{ graphs on } [n] \}$

For $H \in \Omega$

$$P(H) = P(G(n,p) = H) = p^{e(H)} (1-p)^{\binom{n}{2} - e(H)}$$



$G(n,p)$ = this prob. space

$G \in G(n,p)$ or $G = G(n,p)$ or $G \sim G(n,p)$

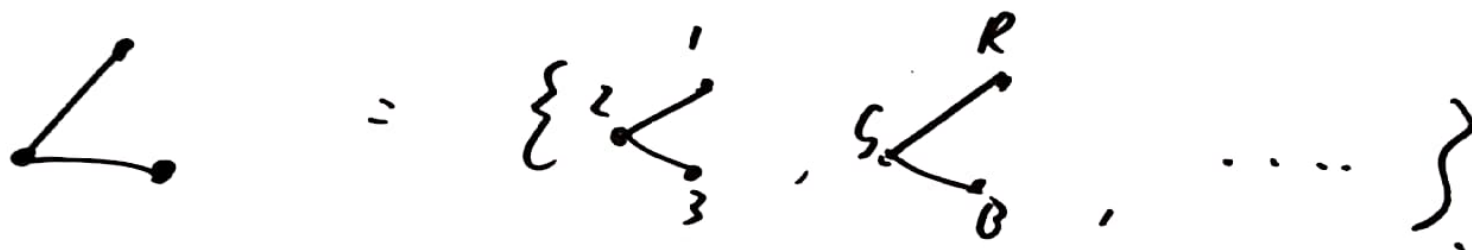
all mean G is a random graph s.t. $\otimes$ holds.

If $p = \frac{1}{2}$ all graphs on $[n]$ equally likely.

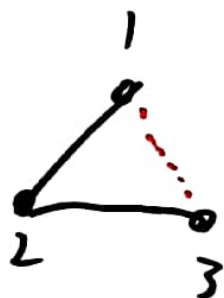
"random labelled graph"

$$G = (V, E)$$

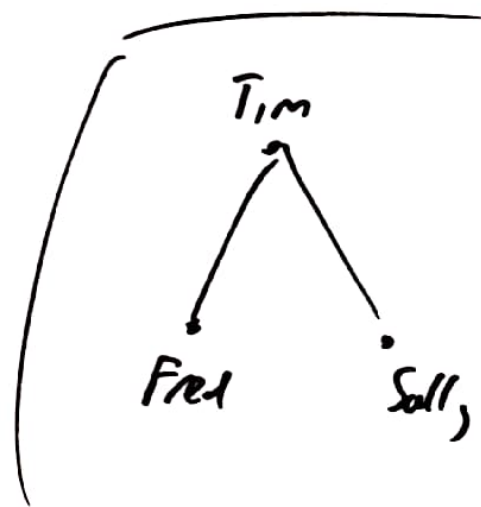
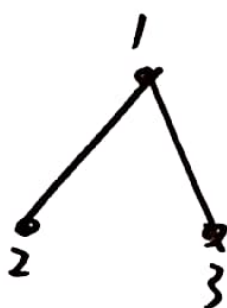
"unlabelled graph" - IM class



$$P(G(3, p) \cong \begin{array}{c} x \\ \swarrow \quad \searrow \\ y \quad z \end{array}) = 3p^2(1-p).$$



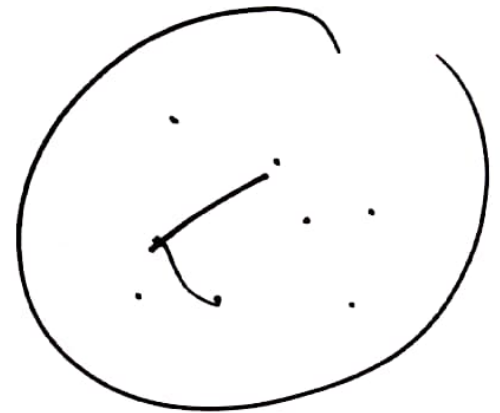
$$p^2(1-p)$$



Model $G(n, p)$ often called the binomial random graph model

$$e(G(n, p)) \sim B_i(N, p) \quad N = \binom{n}{2}$$

Sometimes called Erdős-Rényi model



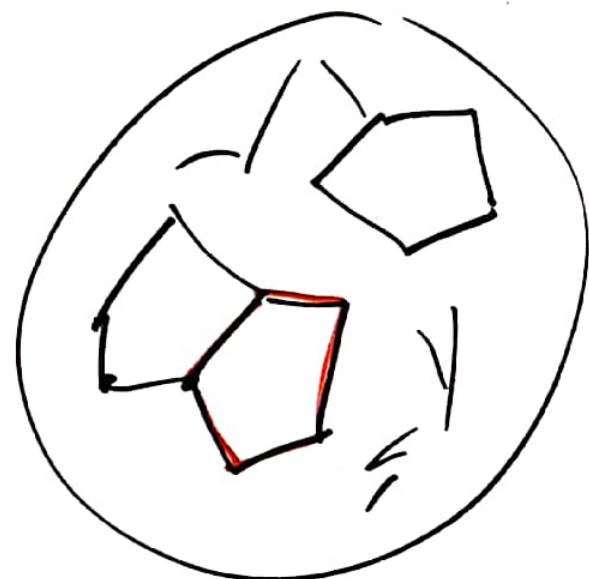
Expected # of cycles of a given length. in $G(n, p)$.

Fix $r \geq 3$, let $X_r = \#$ r -cycles in $G(n, p)$

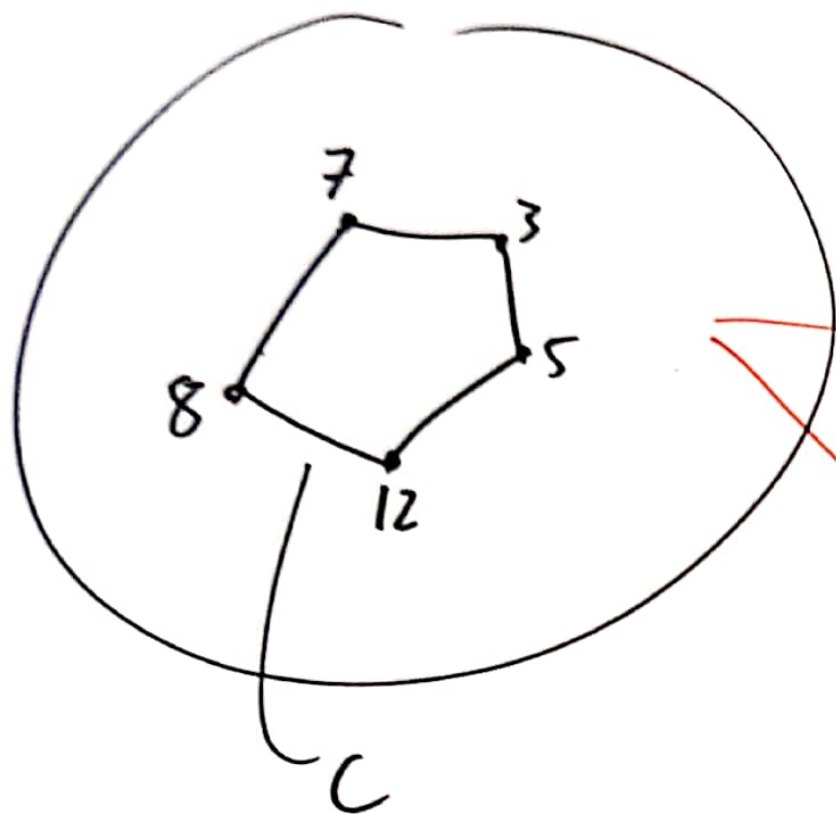
X_r counts how many of the events A_i hold where

$A_i = \{i^{\text{th}} \text{ possible cycle is present}\}$.

$$E[X_r] = \sum_i \underbrace{P(A_i)}_{p^r} \quad \text{cycle in } K_n \text{ of length } r$$



$$V = [n] = V(K_n)$$



Can describe aK by listing
vertices

eg. $(7, 3, 5, 12, 8)$

distinct

$(12, 5, 3, 7, 8)$

2r lists

There are $n(n-1)(n-2) \cdots (n-r+1)$ lists of r distinct vxs.

$\therefore$ there are $\frac{n(n-1) \cdots (n-r+1)}{2r}$ r -cycles in K_n

$$|E[X_r]| = \frac{2r}{p^r}.$$

Lemma 1.8 (Markov's Inequality). Let X be a RV taking non-negative values, and $t > 0$ a constant.

$$\text{Then } P(X \geq t) \leq \frac{E[X]}{t}.$$

Proof $X \geq t I_{\{X \geq t\}}$ always holds.

$$\therefore E[X] \geq E[\quad] = t P(X \geq t) \quad \square$$

Notation / definitions

$\chi(G)$ = chromatic number

= $\min \{k : G \text{ has a proper vertex colouring using } k \text{ colours}\}$

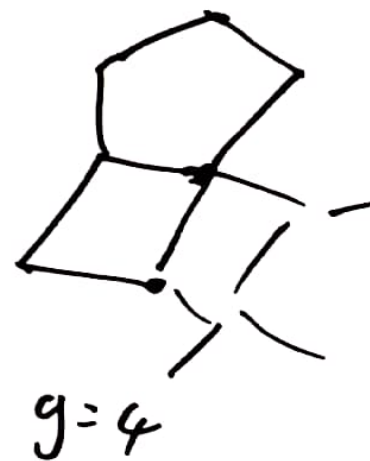
$\alpha(G)$ = independence number

= size of largest of independent set

$g(G)$ = girth of G . = length of shortest cycle

or ∞ if no cycle

$$\forall G: \chi(G) \geq \frac{|G|}{\alpha(G)}$$



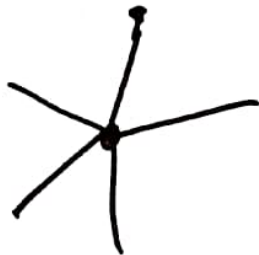
$g(G) \geq l \Rightarrow$ no cycles in G of length $3, 4, \dots, l-1$.

Theorem 1.9 (Erdős 1959).

For any k and l there exists G with $\chi(G) \geq k$
and $g(G) \geq l$.



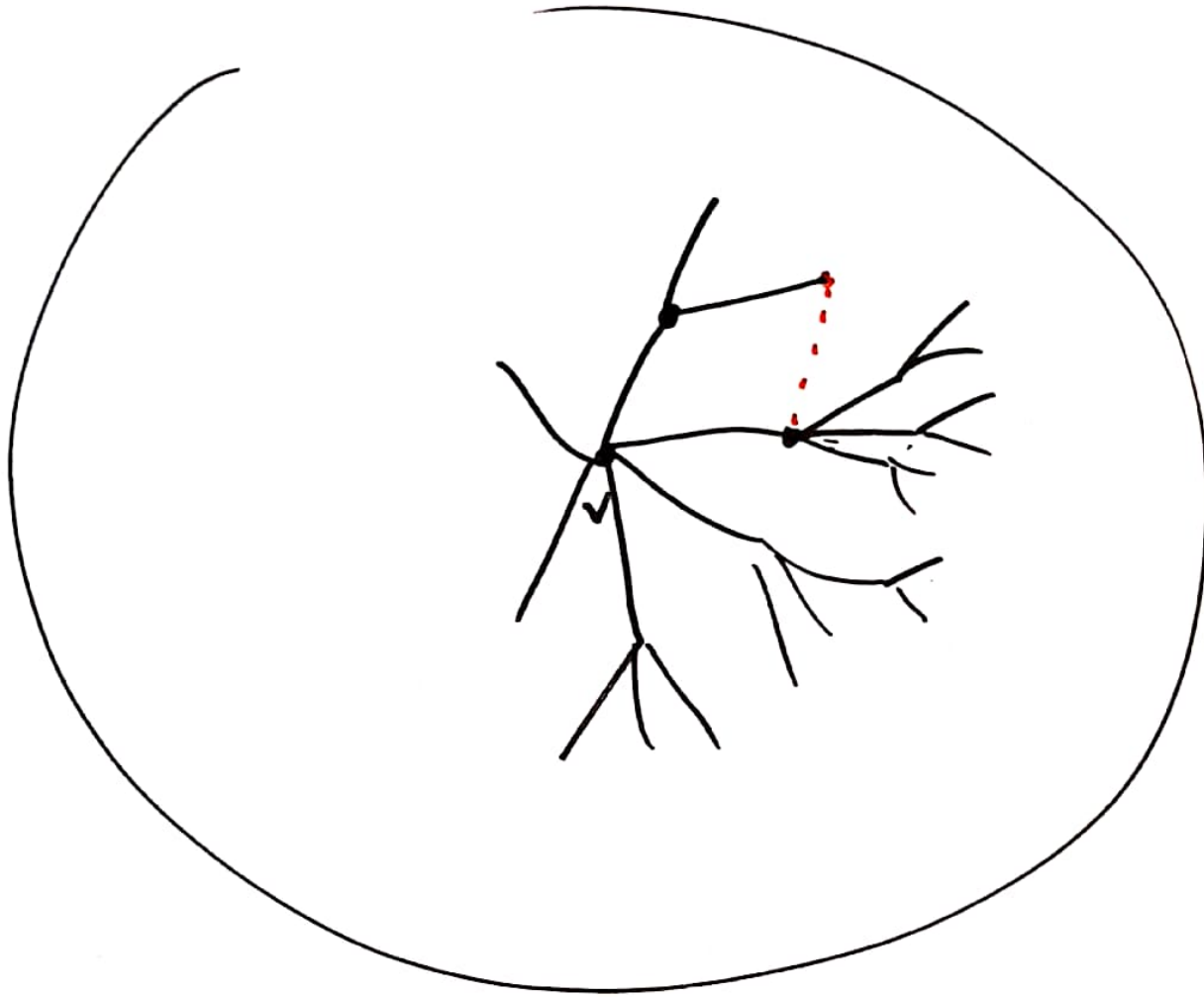
K_{10}



$g(G) \geq 20$

No 3-cycle, no Δ

$$g(4) \geq 20$$



Theorem 1.9 (Erdős 1959)

For any $k, l \quad \exists G$ st $\chi(G) \geq k, \quad g(G) \geq l.$

Idea Consider $G(n, p)$

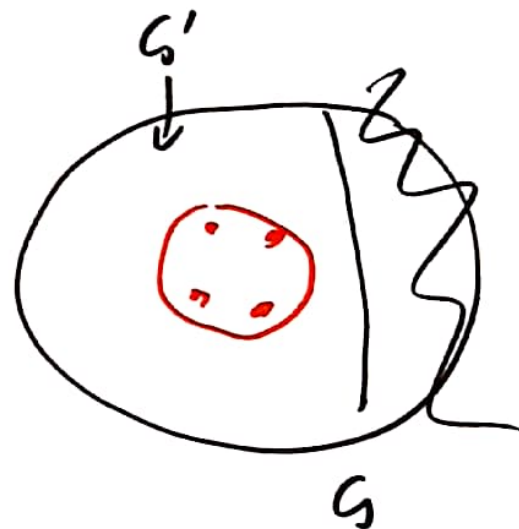
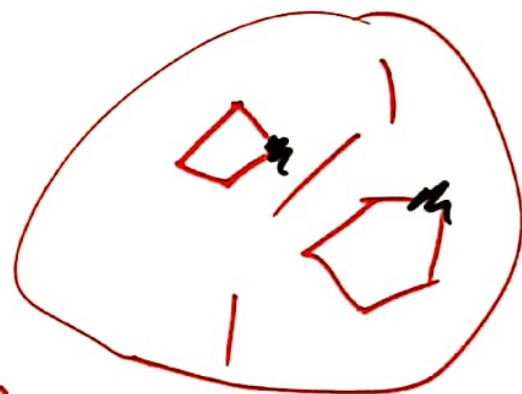
Show separately

(i) Very likely $G(n, p)$ has ~~no~~ short cycles FEW ($< \frac{n}{2}$) (need $p \ll \frac{1}{n}$)

(ii) very likely $G(n, p)$ has ~~large χ~~ .

Small α

$$\alpha(G') \leq \alpha(G)$$



Theorem 1.9 (Erdős 1959) $\forall k, l \exists G : x(G) \geq k, g(G) \geq l.$

Pf Fix $k, l \geq 3$.

Let $p = p(n) = n^{\frac{1}{l}-1}$

(i) Let $X = \#$ short cycles in $G(n, p)$

Then $X = \sum_{r=3}^{l-1} X_r$, $X_r = \#$ r -cycles.

$$(np)^{l-1} \ll n$$

$$np = n^{\frac{1}{l}}$$

$$p = n^{\frac{1}{l}-1}$$

$$\text{so } E[X] = \sum_{r=3}^{l-1} E[X_r] = \sum_{r=3}^{l-1} \frac{n(n-1) \cdots (n-r+1)}{2r} p^r$$

$$\leq \sum_{r=3}^{l-1} n^r p^r = \sum_{r=3}^{l-1} n^{\frac{r}{l}} = O(n^{\frac{l-1}{l}}) = o(n)$$

$$\therefore \text{By Markov } P(X \geq \frac{n}{2}) \leq \frac{E[X]}{n/2} \rightarrow 0$$

(ii) Let $m = m(n) = \lfloor n^{1-\frac{1}{2\epsilon}} \rfloor$

Consider $\mathcal{Y} = \#$ indep. sets in $G(n, p)$ of size exactly m

$$\mathbb{E}[\mathcal{Y}] = \binom{n}{m} (1-p)^{\binom{m}{2}}$$

$$\leq \left(\frac{en}{m} \right)^m e^{-p \binom{m}{2}} = \left(\frac{en}{m} \underbrace{e^{-\frac{p(m-1)}{2}}}_{< 1} \right)^m$$

$\downarrow > 10 \log n$
 n^{-10}

Now $\frac{p(m-1)}{2} \sim \frac{pm}{2} \sim \frac{n^{\frac{1}{\epsilon}-1} n^{1-\frac{1}{2\epsilon}}}{2}$
 $\sim \frac{n^{\frac{1}{2\epsilon}}}{2} > 20 \log n$ if n large.



$$1-x \leq e^{-x}$$

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

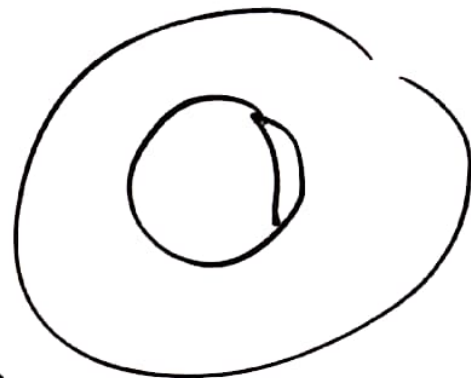
$$\frac{pm}{2} = \frac{n^{\frac{1}{\epsilon}-1} n^{1-\frac{1}{2\epsilon}}}{2} > 10 \log n$$

n^ϵ

If n large enough $\frac{p(m-1)}{2} > 10 \log n \therefore \mathbb{E}[\mathcal{Y}] \leq \left(\frac{en}{m} n^{-10} \right)^m \rightarrow 0$

∴ By Markov
$$P(Y \geq 1) \leq \frac{E[Y]}{1} \rightarrow 0$$

$$P(\alpha(G) \geq m) \rightarrow 0$$



By union bound

$$P(X \geq \frac{n}{2} \text{ or } \alpha(G(n,p)) \geq m) \rightarrow 0$$

$$P(X < \frac{n}{2} \text{ AND } \alpha(G(n,p)) < m) \rightarrow 1.$$

If n large enough

$$\exists \text{ graph } G \text{ s.t. } \begin{cases} G \text{ contains } < \frac{n}{2} \text{ short cycles} \\ \alpha(G) < m. \end{cases}$$

Form G^* by deleting a vertex in each short cycle of G

Then $g(G^*) \geq L$

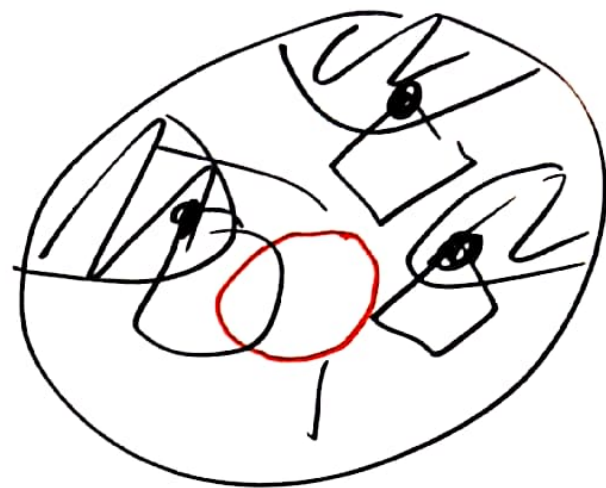
$$|G^*| > \frac{n}{2} \quad (\text{deleted} < \frac{n}{2} \text{ vxs})$$

$$\alpha(G^*) \leq \alpha(G) < m$$

$$\chi(G^*) \geq \frac{|G^*|}{\alpha(G^*)} \geq \frac{n}{2m} \geq \frac{n}{2n^{1-\frac{1}{2k}}} = \frac{n^{\frac{1}{2k}}}{2} > k$$

if n large enough.

□

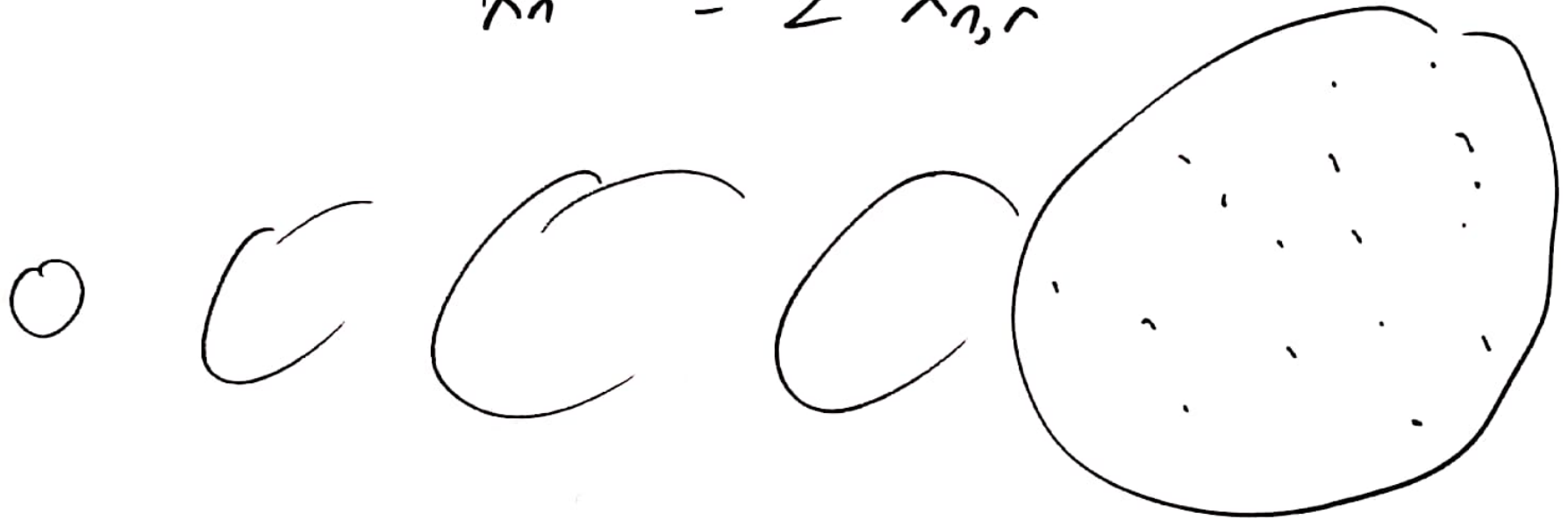


$$\mathbb{E}[Y] \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

$$\text{Had } p = p(n) \quad G_n = G(n, p(n))$$

$$Y_n = \# \dots \dots \dots \text{ in } G_n$$

$$X_n = \sum X_{n,r}$$



§2 The second moment method.

Defn A counting RV is a RV taking non-negative integer values.

Suppose (X_n) is a sequence of counting RVs.

Under like conditions cl. can deduce

(i) • $P(X_n = 0) \rightarrow 1$

(ii) • $P(X_n > 0) \rightarrow 0$.

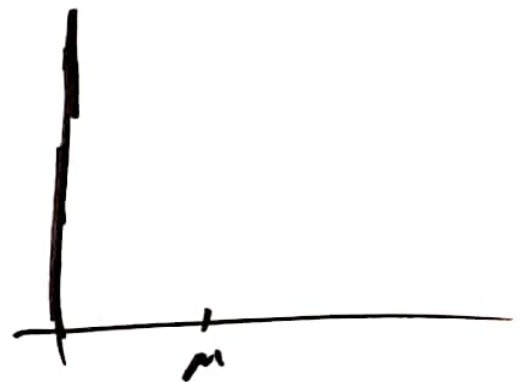
For (i): if $E(X_n) \rightarrow 0$, by Markov

$$P(X_n > 0) = P(X_n \geq 1) \leq \frac{E(X_n)}{1} \rightarrow 0.$$

For (iii) might hope $E[X_n] \rightarrow \infty$ is enough.

eg.
$$X_n = \begin{cases} n^2 & \text{with prob. } \frac{1}{n} \\ 0 & \text{otherwise} \end{cases}$$

$E[X_n] \rightarrow 0$ but $P(X_n > 0) = \frac{1}{n} \rightarrow 0.$



Defn The variance $\text{Var}[X]$ of a RV X is

$$\text{Var}[X] = E[(X - EX)^2] = E[X^2] - (EX)^2$$

Lemma 2.1 Chebyshev's Ineq.

Let X be a RV, $t > 0$. Then $IP(|X - EX| \geq t) \leq \frac{\text{Var}[X]}{t^2}$

Proof Apply Markov to $Y = (X - EX)^2$ $\square$

Cor 2.2 Let (X_n) be a sequence of RVs with $\mu_n = E[X_n] > 0$.

If $\text{Var}[X_n] = o(\mu_n^2)$ then $IP(X_n = 0) \rightarrow 0$

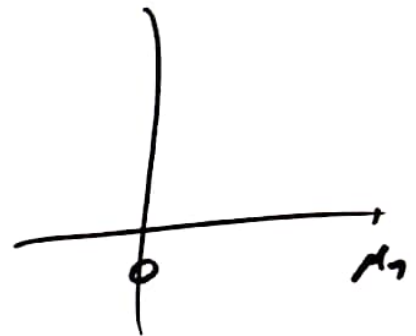
$$IP(X_n > 0) \rightarrow 1.$$

Pf

$$IP(X_n = 0) \leq IP(|X_n - \mu_n| \geq \varepsilon \mu_n)$$

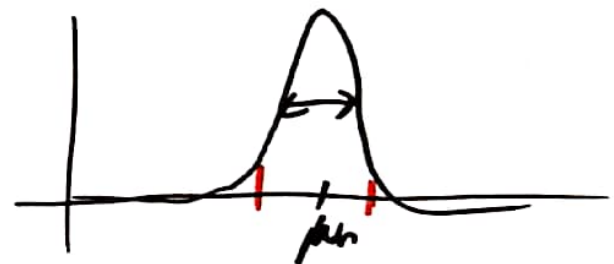
$$\leq \frac{\text{Var}(X_n)}{\varepsilon^2 \mu_n^2} \rightarrow 0.$$

□



Same conditions

$$IP((1-\varepsilon)\mu_n \leq X_n \leq (1+\varepsilon)\mu_n) \rightarrow 1.$$



$$\text{Var } X = E(X^2) - (EX)^2$$

$$\text{Var } [X_n] = o(\mu_n^2) \Leftrightarrow \underbrace{E(X_n^2)}_{\text{second moment}} \sim \mu_n^2$$

Suppose $X = \# A_i$ that hold

$$\text{i.e. } X = \sum_{i=1}^k I_i \quad \text{where } I_i = I_{A_i}$$

$$E[X] = \sum P(A_i)$$

$$E[X^2] = E\left[\sum_i I_i \sum_j I_j\right] = E\left[\sum_i \sum_j I_i I_j\right]$$

$$= \sum_i \sum_j E[I_i I_j] = \sum_{i=1}^k \sum_{j=1}^k P(A_i \cap A_j)$$

1 x 1

$i=1 \quad j=1$
 $i=2 \quad j=2$
 $i=2 \quad j=1$

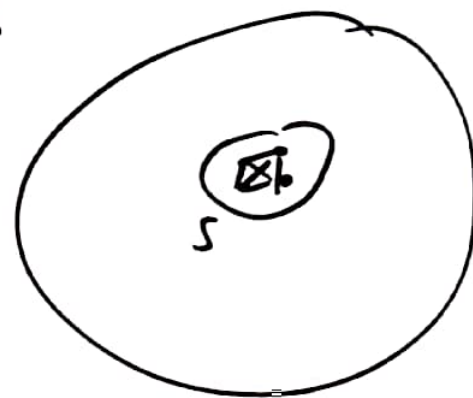
K_4 s in $G(n, p)$

Theorem 2.3 Let $p = p(n)$ be a function of n .

(i) If $n^{2/3}p \rightarrow 0$, then $IP(G(n, p) \text{ contains a } K_4) \rightarrow 0$

(ii) If $n^{2/3}p \rightarrow \infty$, then $IP(\quad \quad \quad) \rightarrow 1$.

Proof Let $X (= X_n)$ be the RV counting # of copies of K_4 in $G(n, p)$. For each $S \subseteq [n]$ with $|S|=4$ let A_S be the event that S induces a K_4 in G .



$$E[X] = \sum_S IP(A_S) = \binom{n}{4} p^6 = \frac{n(n-1)(n-2)(n-3)}{4!} p^6 \sim \frac{n^4 p^6}{4!}$$

$$\mu (= \mu_n) := E(X) \sim \frac{n^4 p^6}{4!}, \quad \mu = O(n^4 p^6)$$

$$n^4 p^6 = (n^{2/3} p)^6 \quad \left[n^{2/3} p \rightarrow 0 \Leftrightarrow \frac{p}{n^{-2/3}} \rightarrow 0 \Leftrightarrow p = o(n^{-2/3}) \right]$$

(i) by assumption $n^{2/3} p \rightarrow 0 \therefore \mu \rightarrow 0$

$$\therefore P(X > 0) = P(X \geq 1) \rightarrow 0 \quad \checkmark$$

(ii) $n^{2/3} p \rightarrow \infty \therefore \boxed{\mu \rightarrow \infty}$

Aim show $\text{Var}(X) = o(\mu^2)$ or, equiv $\boxed{E[X^2] \sim \mu^2}$

$$E[X^2] = \sum_S \sum_T P(A_S \cap A_T) \quad \underline{\text{AIM}} \quad \Sigma \Sigma \sim \mu^2$$

$|S \cap T|$ contribution to $\Sigma \Sigma$

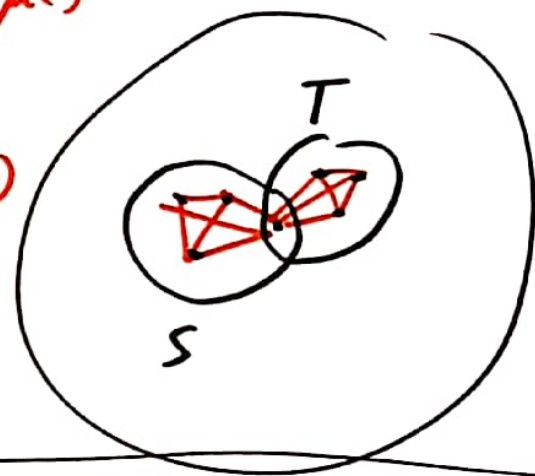
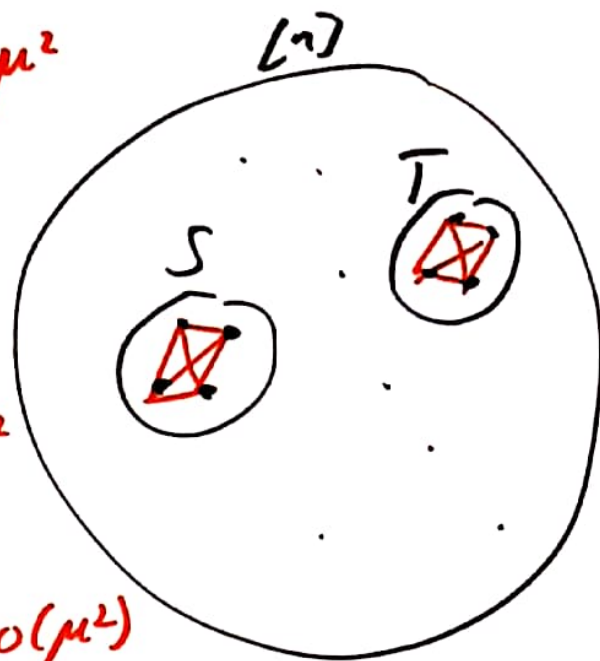
$$0 \quad \binom{n}{4} \binom{n-4}{4} p^{12} \sim \frac{n^4}{4!} \cdot \frac{n^4}{4!} p^{12} \sim \mu^2$$

$$1 \quad \binom{n}{4} 4 \binom{n-4}{3} p^{12} = \Theta(n^7 p^{12}) = o(\mu^2)$$

$$o(\mu^2) \left[2 \quad \binom{n}{4} \binom{4}{2} \binom{n-4}{2} p^{11} = \Theta(n^6 p^{11}) = o(\mu^2) \right.$$

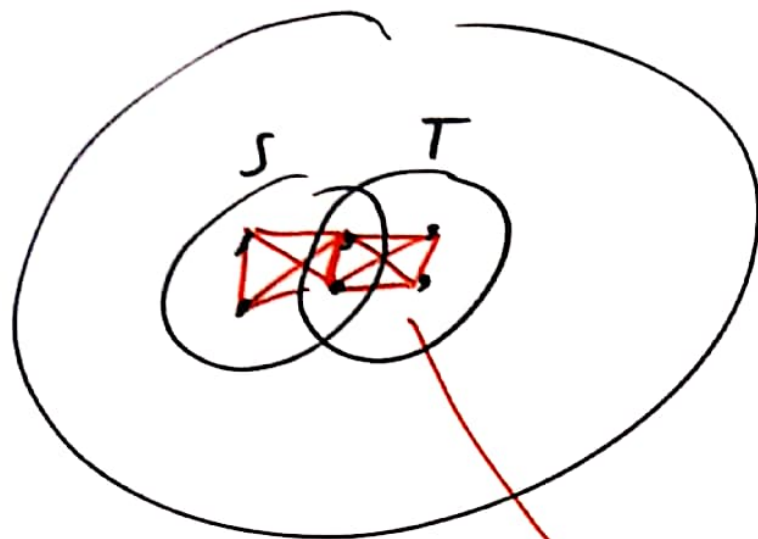
$$3 \quad \binom{n}{4} \binom{4}{3} \binom{n-4}{1} p^9 = \Theta(n^5 p^9) = o(\mu^2)$$

$$4 \quad \binom{n}{4} p^6 = \mu = \mu = o(\mu^2)$$



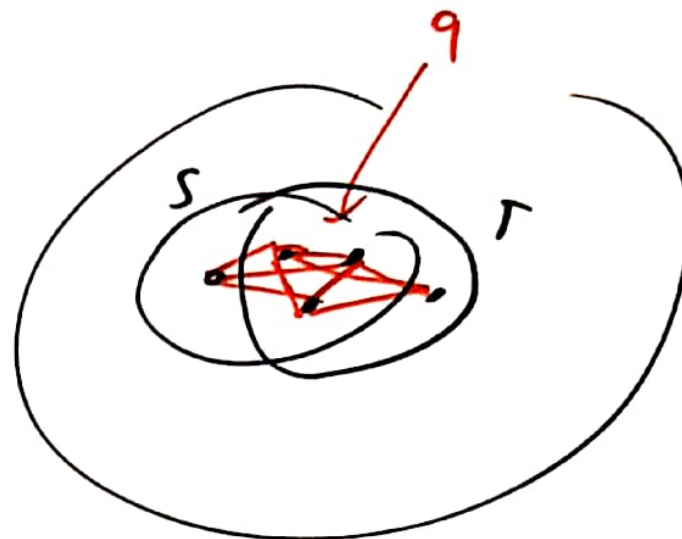
Recall $\mu = \Theta(n^4 p^6) \rightarrow \infty$

$$\begin{aligned} \therefore E[X^2] &\sim \mu^2 \quad / \quad \text{Var}(X) = o(\mu^2) \\ \therefore \text{by Cor 2.2} \quad P(X > 0) &\rightarrow 1. \quad \square \end{aligned}$$

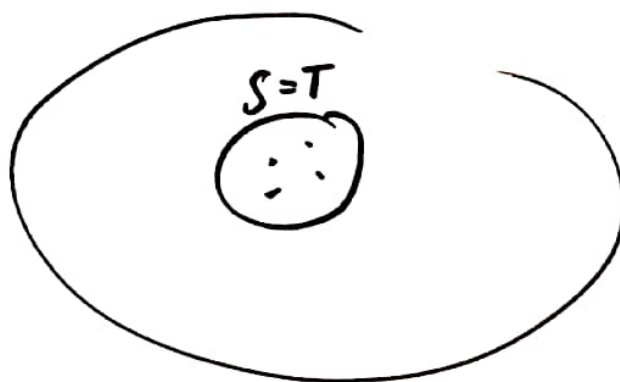


11 edges

$$\binom{n}{4} \binom{4}{2} \binom{n-4}{2} p^{11}$$



$$\binom{n}{4} \binom{4}{3} \binom{n-4}{1} p^9$$



$$\binom{n}{4} p^6 = \mu$$

$$A_S \cap A_S = A_S$$

$$\frac{n^6 p^{11}}{n^8 p^{12}} = \frac{1}{n^2 p} = \frac{1}{\underset{\rightarrow \infty}{n^{4/3}} (\underset{\rightarrow \infty}{n^{2/3} p})} \rightarrow 0$$

$$\frac{n^5 p^9}{n^8 p^{12}} = \frac{1}{n^3 p^3} = \frac{1}{\underset{\rightarrow \infty}{n} (\underset{\rightarrow \infty}{n^{2/3} p})^3} \rightarrow 0$$

Let P be a property of graphs (eg is connected).

A function $p^*(n)$ is a threshold function for P in the model $G(n, p)$ if

$$\bullet \frac{p(n)}{p^*(n)} \rightarrow 0 \Rightarrow \mathbb{P}(G(n, p(n)) \text{ has } P) \rightarrow 0$$

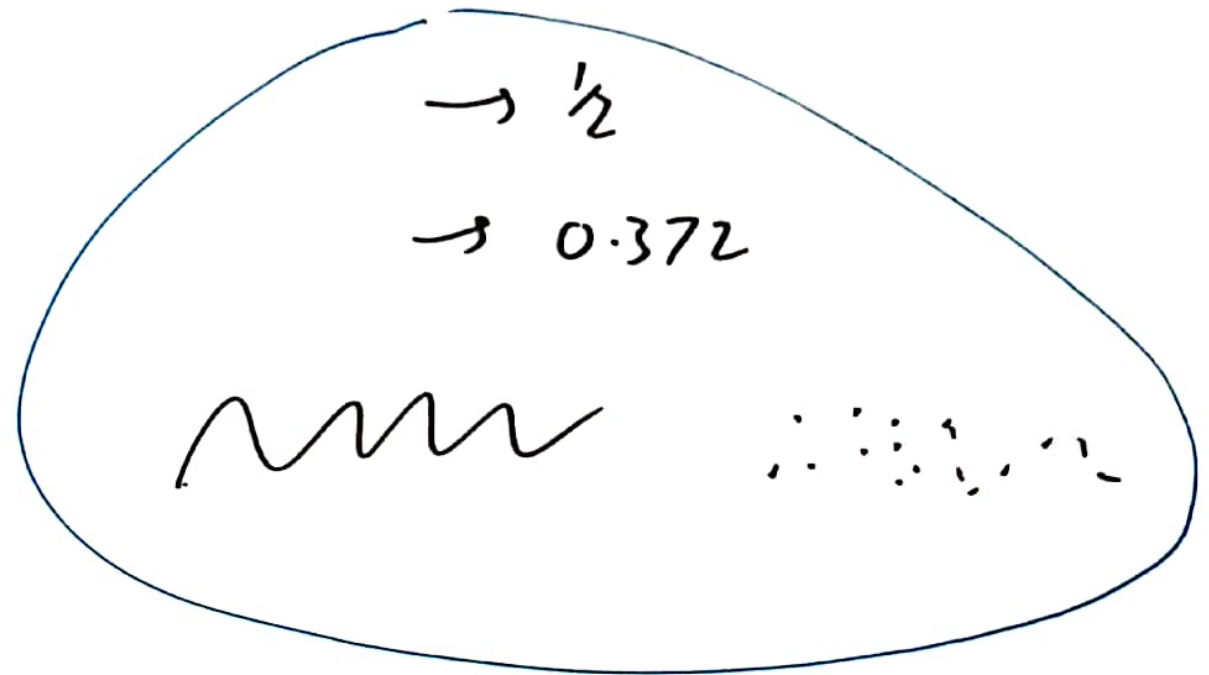
$$\& \bullet \frac{p(n)}{p^*(n)} \rightarrow \infty \Rightarrow \mathbb{P}(G(n, p(n)) \text{ has } P) \rightarrow 1.$$

Theorem 2.3 says $p^*(n) = n^{-2/3}$ is a threshold
for 'contains a K_4 '.

Not unique eg $10n^{-2/3}$ also works

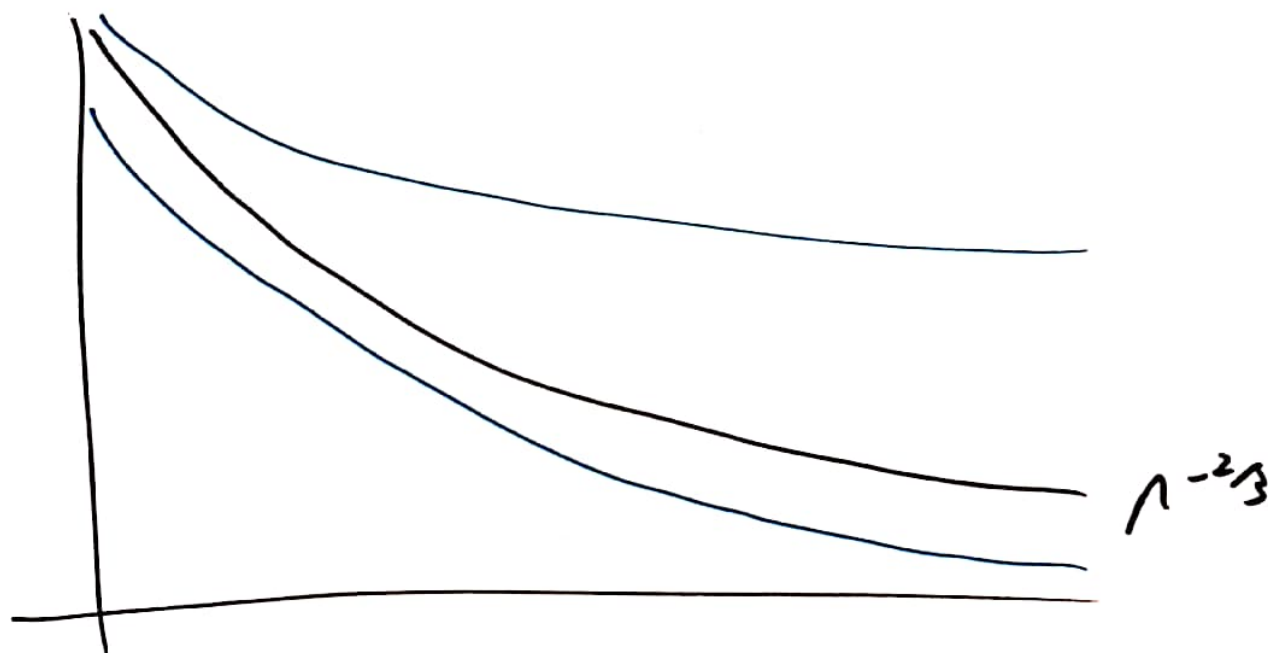
$$q(n) = \mathbb{P}(G(n, p(n)) \text{ has } P) \rightarrow 0$$

$$\rightarrow 1$$



For 'sensible $p(n)$ ' get $\rightarrow 0$ or $\rightarrow 1$
 unless $\Theta(n^{-2/3})$

$$n^{-\alpha} (\log n)^{\beta} (\log \log n)^{\gamma} e^{\delta(\sqrt{\log n})} \dots$$



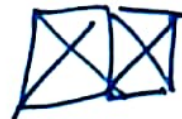
$$\frac{P}{n^{-2/3}} \rightarrow 0$$

$$\frac{P}{n^{-2/3}} \rightarrow \infty$$

"Monotone properties"

$$X = \# A_i \text{ hold}, \quad X = \sum_{i=1}^k I_i \quad I_i = I_{A_i}$$

$$\text{Var}[X] = E(X^2) - (EX)^2$$



$$= \sum_i \sum_j (P(A_i \cap A_j) - P(A_i)P(A_j))$$

Write $i \sim j$ if $i \neq j$ and A_i and A_j are dependent

$$= \sum_i (P(A_i) - P(A_i)^2) + \sum_i \sum_{j \sim i} (P(A_i \cap A_j) - P(A_i)P(A_j)) \quad \begin{matrix} i=j \\ i \sim j \end{matrix}$$

$$\leq \underbrace{\sum_i P(A_i)}_{E[X]} + \underbrace{\sum_i \sum_{j \sim i} P(A_i \cap A_j)}_{\Delta}$$

$i \neq j, i \not\sim j$ A_i, A_j
independent

Cor 2.4 If $\mu := E(X) \rightarrow \infty$ and $\Delta = \sum_i \sum_{j \neq i} P(A_i \cap A_j) = o(\mu^2)$

then $IP(X > 0) \rightarrow 1$.

Proof $Var[X] \leq \mu + \Delta$

$$\therefore \frac{Var(X)}{\mu^2} \leq \frac{1}{\mu} + \frac{\Delta}{\mu^2} \rightarrow 0$$

$$IP(X = 0) \rightarrow 0.$$