

Combinatorics 2017 Fall

week14 note

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Reference:

Extremal Combinatorics with applications in Computer Science. 2nd Edition. Stasys Jukna, Springer.

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Theorem1 Let G be a graph on n vertices such that every vertex has at least one neighbor and at most d non-neighbors. Then $cc(G) = O(d^2 \ln n)$.

Proof: consider the following way for choosing a clique of $G(V, E)$. First, pick every vertex $v \in V$ independently with probability $p = \frac{1}{d+1}$, to get a set $W \subset V$. Then remove from W all vertices having at least one non-neighbor in W . Then we get a clique of G . Applying this way independently t times to get t cliques H_1, \dots, H_t of G , where t is to be determined later. Let X be the number of edges of G not covered by any H_i , and for \forall edge $u w \in E(G)$, let X_{uw} be the indicator random variable of the event "none of H_i covers uw ". Then $X = \sum_{u w} X_{uw}$. Note that if H_i covers uw , if both uw are chosen and none of their \leq neighbors are chosen. Hence

$$\begin{aligned} Pr[H_i \text{ cover } uw] &\geq p^2(1-p)^{2d} = \frac{1}{(1+d)^2} \left(1 - \frac{1}{1+d}\right)^{2d} \\ &\geq \frac{1}{(1+d)^2} [e^{-\frac{1}{1+d} - \frac{1}{2(1+d)^2}}]^{2d} \geq \frac{1}{e^2(1+d)^2} \end{aligned}$$

Since H_i are chosen independently.

$$Pr[\text{none of } H_i \text{ cover } uw] = \prod_{i \in [t]} Pr[H_i \text{ does not cover } uw] \leq \left(1 - \frac{1}{e^2(1+d)^2}\right)^t \leq e^{-\frac{t}{e^2(1+d)^2}}$$

Then by linearity of expectation.

$$E[X] = \sum_{u w} E[X_{uw}] = \sum_{u w} Pr[\text{none of } H_i \text{ cover } uw] \leq \binom{n}{2} e^{-\frac{t}{e^2(1+d)^2}}$$

Take $t = \lfloor cd^2 \ln n \rfloor$ for sufficiently large constant c , we have $E[X] < 1$. Hence, there is at least one choice of t cliques that form a covering of G and $cc(G) \leq t$. \square

Markov's Inequality

Let $X \geq 0$ be a random variable and $t > 0$, then $p(X \geq t) \leq \frac{E[X]}{t}$.

Proof: Let $I_{\{X \geq a\}}$ be the indicator random variable which have value 1 if $X \geq a$ and 0 otherwise. Then $aI_{\{X \geq a\}} \leq X$. In fact if $X < a$, $I_{\{X \geq a\}} = 0$ and so $aI_{\{X \geq a\}} = 0 \leq X$. If $X \geq a$, $I_{\{X \geq a\}} = 1$ and so $aI_{\{X \geq a\}} = a \leq X$. Since E is a monotonically increasing function, taking expectation of both sides, we have $E[aI_{\{X \geq a\}}] \leq E[X]$. Since $E[aI_{\{X \geq a\}}] = aE[I_{\{X \geq a\}}] = aPr[X \geq a] \leq E[X]$. \square

Definition: The random graph $G(n, p)$ for $0 \leq p \leq 1$ is a graph with vertex set $[n]$, where each of potential $\binom{n}{2}$ edges appears with probability p independent of other edges.

Corollary: Let $X_n \geq 0$ be integer value random variable in (Ω_n, P_n) , $n \in \mathbb{Z}_{\geq 0}$. If $E[X_n] \rightarrow 0$ as $n \rightarrow +\infty$, then $Pr[X_n = 0] \rightarrow 1$ as $n \rightarrow +\infty$, we say $X_n = 0$ almost surely occur.

Proof: Let $t = 1$, $Pr[X_n \geq 1] \leq \frac{E[X_n]}{1} \rightarrow 0$ as $n \rightarrow +\infty$. \square

Theorem For a random graph $G(n, p)$ for some fixed $p \in (0, 1)$. then $Pr[\alpha(G) \leq \left\lceil \frac{2 \ln n}{p} \right\rceil] \rightarrow 1$ as $n \rightarrow \infty$.

Proof: Let $k = \left\lceil \frac{2 \ln n}{p} \right\rceil$. For $\forall S \in \binom{[n]}{k+1}$, Let A_S be the event that S is an independent set, Let $X_n = \sum_{S \in \binom{[n]}{k+1}} I_{A_S}$ be the number of independent set of size $k+1$, here I_{A_S} be the indicator random variable of the event " A_S occurs". Then

$$\begin{aligned} E[X_n] &= \sum_{S \in \binom{[n]}{k+1}} E[I_{A_S}] = \sum_{S \in \binom{[n]}{k+1}} Pr[A_S] \\ &= \binom{n}{k+1} (1-p)^{\binom{k+1}{2}} \leq \frac{n^{k+1}}{(k+1)!} e^{-p \binom{k+1}{2}} \\ &= \frac{1}{(k+1)!} (ne^{-p \frac{k}{2}})^{k+1} \leq \frac{1}{(k+1)!} \rightarrow 0 \end{aligned}$$

Then $Pr[X_n = 0] \rightarrow 1$ as $n \rightarrow \infty \implies Pr[\alpha(G) \leq \left\lceil \frac{2 \ln n}{p} \right\rceil] \rightarrow 1$. \square

Definition: The girth of G denoted by $g(G)$ is the length of a shortest cycle of G .

Recall: $\chi(G) \geq n = V(G)$.

Theorem For any $k \in \mathbb{N}^+$, there exists a graph G with $\chi(G) \geq k$ and $g(G) \geq k$.

Proof: Consider $G(n, p)$ where p will be determined later. let X = be the number cycles of length less than k in G , and X_i be the number of cycles of length i , $i \in [k-1]$. Then

$$E[X] = \sum_{i=3}^{k-1} E[X_i] = \sum_{i=3}^{k-1} \frac{n(n-1) \cdots (n-i+1)}{2i} p^i.$$

where $\frac{n(n-1) \cdots (n-i+1)}{2i}$ is the number of of cycles of length i in K_n . Then $E[X] \leq \sum_{i=3}^{k-1} \frac{n^i p^i}{2i} \leq$

$\sum_{i=0}^{k-1} (np)^i = \frac{(np)^k - 1}{np - 1}$. By Markov's Inequality

$$Pr[X \geq \frac{n}{2}] \leq \frac{E[X]}{n/2} \leq \frac{2[(np)^k - 1]}{n(np - 1)}.$$

Let $p = n^{-\frac{k-1}{k}}$, $\implies Pr[X \geq \frac{n}{2}] \leq \frac{2(n-1)}{n(n^{\frac{1}{k}} - 1)} \rightarrow 0$ as $n \rightarrow +\infty$. Let $t = \left\lceil \frac{2 \ln n}{p} \right\rceil \leq 3 \ln n \cdot n^{\frac{k-1}{k}}$, recall $\alpha(G) \leq t$ almost surely. Then $\exists G$ on n vertices, $\alpha(G) \leq t$ and with $\leq \frac{n}{2}$ cycles of length less than k . By deleting from each cycle of length less than k , we have a graph $G' \subset G$, with No cycles of length less than k , and $|V(G')| \geq n - \frac{n}{2} = \frac{n}{2}$. $\alpha(G') \leq \alpha(G) \leq 3 \ln n \cdot n^{\frac{k-1}{k}}$. So $\chi(G') \geq \frac{|V(G')|}{\alpha(G')} \geq \frac{n/2}{3 \ln n \cdot n^{\frac{k-1}{k}}} = \frac{n^{\frac{1}{k}}}{6 \ln n} \gg k$, where n is sufficiently large. \square

2017/12/29

The Lovasz Sieve

Recall union bound $Pr[A \cup B] \leq Pr[A] + Pr[B]$. Let B_1, \dots, B_m be bad events, and $A = \overline{B_1} \cap \dots \cap \overline{B_m}$, then

$$Pr[A] = Pr[\overline{B_1} \cap \dots \cap \overline{B_m}] = Pr[\overline{B_1 \cup \dots \cup B_m}] = 1 - Pr[B_1 \cup \dots \cup B_m] \geq 1 - \sum_{i=1}^m Pr[B_i].$$

- If $\sum_{i=1}^m Pr[B_i] < 1$, then $Pr[A] > 0$.
- If B_1, \dots, B_m are independent, then

$$Pr[A] = Pr[\overline{B_1} \cap \dots \cap \overline{B_m}] = \prod_{i=1}^m Pr[\overline{B_i}] = \prod_{i=1}^m (1 - Pr[B_i])$$

So if $Pr[B_i] < 1, i \in [m]$, then $Pr[A] > 0$.

- What if some B_i, B_j are not independent?

Conditional probability: $Pr[A|B] = Pr[A \cap B] / Pr[B]$.

A and B are independent iff $Pr[A \cap B] = Pr[A]Pr[B] \iff Pr[A|B] = Pr[A]$.

$$Pr[A|(B \cap C)] = \frac{Pr[A \cap B \cap C]}{Pr[B \cap C]} = \frac{Pr[A \cap B \cap C]}{Pr[C]} \cdot \frac{Pr[B \cap C]}{Pr[C]} = Pr[(A \cap B)|C] / Pr[B|C].$$

$$Pr[A|(B \cap C)] \cdot Pr[B|C] \cdot Pr[C] = \frac{Pr[A \cap B \cap C]}{Pr[B \cap C]} \cdot \frac{Pr[B \cap C]}{Pr[C]} \cdot Pr[C] = Pr[A \cap B \cap C].$$

Definition: An event A is mutually independent of events B_1, \dots, B_m if $Pr[A|(C_1 \cap \dots \cap C_m)] = Pr[A], C_i \in \{B_i, \overline{B_i}\}, i \in [m]$.

Note: The fact A is mutually independent with each of B_1, \dots, B_m does not mean that A is mutually independent of B_1, \dots, B_m .

Eg: consider flipping a fair coin twice, let $B_i, i = 1, 2$ be the event that the i-th flip is a head,

and let A be the event that both flips are the same. Then A is independent with each B_i . But $Pr[A|B_1 \cap B_2] = 1$.

Definition: Let A_1, \dots, A_n be events, $G = ([n], E)$ is called a dependency graph if $\forall i, A_i$ is mutually independent of all events A_j with $j \in [n] \setminus \{i\}$ and $\{i, j\} \notin E$. The smallest degree of among all such graphs G is called the degree of dependency of A_1, \dots, A_n .

Lemma: Let A_1, \dots, A_n be events with $Pr[A_i] \leq p < 1, \forall i$, and let d be the degree of dependency. If $4pd \leq 1$, then $Pr[\overline{A_1} \cap \dots \cap \overline{A_n}] > 0$.

Proof: Fix a dependency graph G of A_1, \dots, A_n of degree d . Prove by induction on m or $S, S \subset [n], |S| = s < n, \forall i \notin S, Pr[A_i | \cap_{i \in S} \overline{A_j}] \leq 2p$ that (Claim) for any m events of A_1, \dots, A_n , say A_1, \dots, A_m for convenience, $Pr[A_1 | \overline{A_2} \cap \dots \cap \overline{A_m}] \leq 2p$. $m = 1$ is true. Let $2, \dots, k \in [2, m]$ which are adjacent to 1 in G . Then $Pr[A_1 | \overline{A_2} \cap \dots \cap \overline{A_m}] = \frac{Pr[(A_1 \cap \overline{A_2} \cap \dots \cap \overline{A_k}) | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})]}{Pr[(\overline{A_2} \cap \dots \cap \overline{A_k}) | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})]}$. Since A_1 is mutually independent of A_{k+1}, \dots, A_m , the numerator

$$Pr[(A_1 \cap \overline{A_2} \cap \dots \cap \overline{A_k}) | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})] \leq Pr[A_1 | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})] = Pr[A_1] \leq p.$$

The denominator

$$\begin{aligned} & Pr[(\overline{A_2} \cap \dots \cap \overline{A_k}) | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})] \\ &= 1 - Pr[(A_2 \cup \dots \cup A_k) | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})] \\ &\geq 1 - \sum_{i=2}^k Pr[A_i | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})] \\ &\geq 1 - (k-1)2p \geq \frac{1}{2} \end{aligned}$$

(since $k-1 \leq d, 2pd \leq \frac{1}{2}$)

Then $Pr[A_1 | \overline{A_2} \cap \dots \cap \overline{A_m}] \leq p / (\frac{1}{2}) = 2p$. Finally,

$$Pr[\overline{A_1} \cap \dots \cap \overline{A_n}] = \prod_{i=1}^n Pr[\overline{A_i} | (\overline{A_{i+1}} \cap \dots \cap \overline{A_n})] = \prod_{i=1}^n (1 - Pr[A_i | (\overline{A_{i+1}} \cap \dots \cap \overline{A_n})]) \geq (1-2p)^n > 0,$$

when A_i 's probability are very different. \square

Lemma: Let $G = (V, E)$ be a dependency graph of event A_1, \dots, A_n . Suppose there exist real numbers $x_1, \dots, x_n, 0 \leq x_i \leq 1$, so that for all $i, Pr[A_i] = x_i \cdot \prod_{\{i,j\} \in E} (1 - x_j)$. Then $Pr[\overline{A_1} \cap \dots \cap \overline{A_n}] \geq \prod_{i=1}^n (1 - x_i) > 0$.

Proof: Similar to the above lemma. Replace the claim by for any m events $Pr[A_1 | \overline{A_2} \cap \dots \cap \overline{A_m}] \leq x_1$, we also prove it by induction on m . Using same notations, we have

$$Pr[A_1 | \overline{A_2} \cap \dots \cap \overline{A_m}] = \frac{Pr[(A_1 \cap \overline{A_2} \cap \dots \cap \overline{A_k}) | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})]}{\prod_{j=2}^k Pr[A_j | (\overline{A_{j+1}} \cap \dots \cap \overline{A_m})]}$$

the numerator $Pr[(A_1 \cap \overline{A_2} \cap \dots \cap \overline{A_k}) | (\overline{A_{k+1}} \cap \dots \cap \overline{A_m})] \leq Pr[A_1] \leq x_1 \cdot \prod_{\{1,j\} \in E} (1 - x_j)$.

the denominator

$$\prod_{j=2}^k Pr[\overline{A_j} | (\overline{A_{j+1}} \cap \dots \cap \overline{A_m})] = \prod_{j=2}^k (1 - Pr[A_j | (\overline{A_{j+1}} \cap \dots \cap \overline{A_m})]) \geq \prod_{j=2}^k (1 - x_j) \geq \prod_{\{1,j\} \in E} (1 - x_j)$$

$\implies Pr[A_1 | (\overline{A_2} \cap \dots \cap \overline{A_m})] \leq x_1$. Then $Pr[\overline{A_1} \cap \dots \cap \overline{A_m}] \geq \prod_{i=1}^n (1 - x_i)$. \square

Corollary: Let A_1, \dots, A_n be events with $Pr[A_i] \leq p$, degree of dependency is d . If $ep(d+1) \leq 1$, then $Pr[\overline{A_1} \cap \dots \cap \overline{A_n}] > 0$.

Proof: Let $x_i = \frac{1}{d+1}$, then

$$x_i(1 - x_i)^d = \frac{1}{d+1} \left(1 - \frac{1}{d+1}\right)^d > \frac{1}{d+1} \frac{1}{e} \geq p \geq Pr[A_i].$$

Since $\ln(1+t) = t - \frac{t^2}{2} + \frac{t^3}{3} - \dots$, $-1 < t \leq 1$. we have

$$\begin{cases} 1 - t > e^{-t - \frac{t^2}{2}}, \text{ if } 0 < t < 1; \\ 1 - t < e^{-t - \frac{t^2}{2}}, \text{ if } -1 < t < 0. \end{cases}$$

Then $(1 - \frac{1}{d+1})^d > e^{-d(\frac{1}{d+1} + \frac{1}{2(d+1)^2})} > e^{-1}$. Or since $(1 - \frac{1}{d+1})^d = (\frac{d}{d+1})^d = \frac{1}{(1 + \frac{1}{d})^d} > \frac{1}{e}$. \square

Recall: Let A_1, \dots, A_n be events, $Pr[A_i] = p, i \in [n]$. If all events are mutually independent, then $Pr[\cup_{i=1}^n A_i] = 1 - Pr[\cap_{i=1}^n \overline{A_i}] = 1 - (1-p)^n \geq 1 - e^{-pn}$. But if not, we have two more cases.

(1) k -wise independent, i.e. $Pr[\cap_{i \in I} A_i] = p^{|I|}$ for all $|I| \leq k$.

(2) k -wise independent, i.e. let $\delta < 1$. $|Pr[\cap_{i \in I} A_i] - p^{|I|}| \leq \delta$ for all $|I| \leq k$.

Recall Inclusion-Exclusion Principle: $|A_1 \cup A_2 \cup \dots \cup A_n| = \sum_{\emptyset \neq I \subseteq [n]} (-1)^{|I|+1} |A_I|$.

Lemma Bonferroni inequality, for each even $k \geq 2$,

$$\sum_{v=1}^k (-1)^{v+1} \sum_{|I|=v} A_I \leq |\cup_{i=1}^n A_i| \leq \sum_{v=1}^{k+1} (-1)^{v+1} \sum_{|I|=v} A_I.$$

Proof: By induction on n , EX.