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Univeristy of South Carolina, Spring, 2019

Introduction



The topic course is mostly based the textbook "The probabilistic Method" by Noga Alon and Joel Spencer (third edition 2008, John Wiley & Sons, Inc. ISBN 9780470170205 or fourth edition ISBN-13: 978-1119061953.)







Selected topics



- Linearity of Expectation (2 weeks)
- Alterations (1 week)
- The second moment method (1 week)
- The Local Lemma (1-2 weeks)
- Correlation Inequalities (1 week)
- Large deviations (1-2 weeks)
- Poisson Paradigm (1 week)
- Random graphs (2 weeks)
- Discrepancy (1 week)
- Entropy (1 week)



Subtopics

Linearity of Expectation (2)

- Disjoint pairs
- *k*-sets
- Balancing vectors
- Unbalancing lights
- Brégman's Theorem
- Hamliton paths
- Independence number
- Turán Theorem



Disjoint pairs



 $\mathcal{F} \subset 2^{[n]}.$ $\mathbf{d}(\mathcal{F}) := |\{(F, F') \colon F, F' \in \mathcal{F}, F \cap F' = \emptyset\}|.$



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Theorem [Alon-Frankl, 1985]: If $|\mathcal{F}| = 2^{(1/2+\delta)n}$, then

 $d(\mathcal{F}) < |\mathcal{F}|^{2-\delta^2/2}.$









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Proof



Let $m := 2^{(1/2+\delta)n}$. Suppose $d(\mathcal{F}) < m^{2-\delta^2/2}$. Pick independently t members A_1, A_2, \ldots, A_t of \mathcal{F} with repetitions at random.

$$\Pr(|\cup_{i=1}^{t} A_i| \le \frac{n}{2})$$

$$\le \sum_{|S|=\frac{n}{2}} \Pr(\wedge_{i=1}^{t} (A_i \subset S))$$

$$\le 2^n \left(\frac{2^{n/2}}{2^{(1/2+\delta)n}}\right)^t$$

$$= 2^{n(1-\delta t)}.$$









Let $v(B) = |\{A \in \mathcal{F} \colon B \cap A = \emptyset\}|$. Then

$$\sum_{B} v(B) = 2d(\mathcal{F}) \ge 2m^{2-\delta^2/2}.$$









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Let Y be a random variable whose value is the number of members $B \in \mathcal{F}$ that is disjoint to all A_i $1 \le i \le t$.









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Let Y be a random variable whose value is the number of members $B \in \mathcal{F}$ that is disjoint to all A_i $1 \le i \le t$. Then

$$E(|Y|) = \sum_{B \in \mathcal{F}} \left(\frac{v(B)}{m}\right)^t$$

$$\geq \frac{1}{m^{t-1}} \left(\frac{\sum_B v(B)}{m}\right)^t$$

$$\geq 2m^{1-t\delta^2/2}.$$









Since $Y \leq m$, we get

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Choose $t = \lfloor 1 + \frac{1}{\delta} \rfloor$. We have $m^{-t\delta^2/2} > 2^{n(1-\delta t)}$.

Thus, with positive probability, $|\bigcup_{i=1}^{t} A_i| > \frac{n}{2}$ and $\bigcup_{i=1}^{t} A_i$ is disjoint to more than $2^{n/2}$ members of \mathcal{F} . Contradiction. \Box



Linearity of expectation

Let X_1, X_2, \ldots, X_n be random variables and $X = \sum_{i=1}^n c_i X_i$. Then

$$\mathbf{E}(X) = \sum_{i=1}^{n} c_i \mathbf{E}(X_i).$$



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Philosophy: There is a point in the probability space for which $X \ge E(X)$ and a point for $X \le E(X)$.







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Let X be the number of crossing edges (from L to R). Let X_{uv} be the indicator variable of the edge uv is crossing.

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$$E(X_{uv}) = \frac{1}{2}.$$
$$E(X) = \sum_{uv \in E} E(X_{uv}) = \frac{m}{2}.$$









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k-sets



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• A k-set F is crossing if it contains precisely one point form each V_i .

Theorem: Suppose h(F) = +1 for all crossing k-sets F. Then there is an $S \subset V$ for which

$$|h(S)| \ge c_k n^k.$$





A Lemma

Lemma: Let P_k be the set of all homogeneous polynomials $f(p_1, \ldots, p_k)$ of degree k with all coefficients have absolute value at most one and $p_1p_2 \cdots p_k$ having coefficient one. Then for all $f \in P_k$ there exists $p_1, \ldots, p_k \in [0, 1]$ with

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Here $c_k > 0$, independent of n.





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Here $c_k > 0$, independent of n.

Proof: Let $M(f) = \max_{p_1,\dots,p_k} |f(p_1,\dots,p_k)|$. Note P_k is compact and M is continuous. M reaches its minimum value c_k at some point f_0 . We have

$$c_k = M(f_0) > 0.$$



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Say F has type (a_1, \dots, a_k) if $|F \cap V_i| = a_i, 1 \le i \le k$. For these F ,
$$E(X_F) = h(F)p_1^{a_1} \cdots p_k^{a_k}.$$







$\mathbf{E}(X) = \sum_{\substack{k\\\sum_{i=1}^{k} a_i = k}} p_1^{a_1} \cdots p_k^{a_k} \sum_{\substack{F \text{ of type } (a_1, \dots, a_k)}} h(F).$





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Let $f(p_1, \dots, p_k) = \frac{1}{n^k} E(X)$. Then $f \in P_k$.
Now select $p_1, \dots, p_k \in [0, 1]$ with $|f(p_1, \dots, p_k)| \ge c_k$.

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There exists a S such that $|h(S)| \ge c_k n^k$.





Balancing vectors

Theorem: Let v_1, \ldots, v_n are n unit vector in \mathbb{R}^n . Then there exist $\epsilon_1, \ldots, \epsilon_n = \pm 1$ so that

$$\|\epsilon_1 v_1 + \dots + \epsilon_n v_n\| \leq \sqrt{n},$$

and also there exist $\epsilon_1, \ldots, \epsilon_n = \pm 1$ so that

$$\|\epsilon_1 v_1 + \dots + \epsilon_n v_n\| \ge \sqrt{n}.$$



Proof



Let $\epsilon_1, \ldots, \epsilon_n$ be selected uniformly and independently from $\{+1, -1\}$. Let $X = \|\epsilon_1 v_1 + \cdots + \epsilon_n v_n\|^2$.



Proof



Let $\epsilon_1, \ldots, \epsilon_n$ be selected uniformly and independently from $\{+1, -1\}$. Let $X = \|\epsilon_1 v_1 + \dots + \epsilon_n v_n\|^2$. $\mathbf{E}(X) = \mathbf{E}(\sum \epsilon_i \epsilon_j v_i \cdot v_j)$ i, j=1 $= \sum \mathbf{E}(\epsilon_i \epsilon_j) v_i \cdot v_j$ i.i=1 $=\sum \delta_i^j v_i \cdot v_j$ i.i=1n $= \sum ||v_i||^2 = n.$ i=1





An extension



Theorem: Let $v_1, \ldots, v_n \in \mathbb{R}^n$, all $||v_i|| \leq 1$. Let $p_1, p_2, \ldots, p_n \in [0, 1]$ be arbitrary and set $w = p_1v_1 + p_2v_2 + \cdots + p_nv_n$. Then there exist $\epsilon_1, \ldots, \epsilon_n \in \{0, 1\}$ so that setting $v = \epsilon_1v_1 + \cdots + \epsilon_nv_n$,

$$\|w - v\| \le \frac{\sqrt{n}}{2}$$





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Hint: Pick ϵ_i independently with

$$\Pr(\epsilon_i = 1) = p_i, \quad \Pr(\epsilon_i = 0) = 1 - p_i.$$

The proof is similar.



Unbalancing lights

Theorem: Let $a_{ij} = \pm 1$ for $1 \le i, j \le n$. Then there exist $x_i, y_j = \pm 1, 1 \le i, j \le n$ so that

$$\sum_{i,j=1}^{n} a_{ij} x_i y_j \ge \left(\sqrt{\frac{2}{\pi}} + o(1)\right) n^{3/2}$$





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Proof: Choose $y_j = 1$ or -1 randomly and independently. Let $R_i = \sum_{i=1}^n a_{ij}y_j$. Let x_i be the sign of R_i . Then

$$\sum_{i,j=1}^{n} a_{ij} x_i y_j = \sum_{i=1}^{n} |R_i|.$$





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Hence,

$$\sum_{i=1}^{n} \mathrm{E}(|R_i|) = \left(\sqrt{\frac{2}{\pi}} + o(1)\right) n^{3/2}.$$



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Brégman's Theorem (1973): $per(A) \leq \prod_{1 \leq i \leq n} (r_i!)^{1/r_i}$.



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Claim: $per(A) \le G(L).$



For any fixed τ . Assume $\tau(1) = 1$. By re-ordering, assume the first row has ones in the first $r := r_1$ columns. For $1 \le j \le r$ let t_j be the permanent of A with the first row and j-th column removed (i.e., $\sigma(1) = j$). Let

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By induction,

$$G(R_2 \cdots R_n | \sigma(1) = j) \ge t_j.$$

$$G(L) \ge \prod_{j=1}^{r} (rt_j)^{t_j/per(A)} = r \prod_{j=1}^{r} (t_j)^{t_j/rt}.$$



Topic Course on Probabilistic Methods (week 2)





Since
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Now we calculate G[L] conditional on a fixed σ . By reordering, assume $\sigma(i) = i$ for all i. Note

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$$G(R) = G(\prod_{i=1}^{n} R_i) = \prod_{i=1}^{n} (r_i!)^{1/r_i}.$$



Topic Course on Probabilistic Methods (week 2)



Hamilton paths



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Proof: Let X be the number of Hamiltonian paths in a random tournament. Write $X = \sum_{\sigma \in S_n} X_{\sigma}$. Here X_{σ} is the indicator random variable for σ giving a Hamilton path.

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We have

$$\mathcal{E}(X) = \sum_{\sigma \in S_n} \mathcal{E}(X_{\sigma}) = n! 2^{1-n}.$$

Done!







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$$\frac{1}{2} \le \lim_{n \to \infty} \left(\frac{P(n)}{n!}\right)^{1/n} \le \frac{1}{2^{3/4}}.$$

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This conjecture was proved by Alon in 1990. **Theorem [Alon, 1990]:** $P(n) \leq cn^{3/2} \frac{n!}{2^{n-1}}$.



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 A_T = (a_{ij}): the adjacency matrix of T, where a_{ij} = 1 if
 - $i \rightarrow j$ and 0 otherwise.

$$F(T) = per(A_T) \le \prod_{i=1}^n (r_i!)^{1/r_i}.$$

Here r_i is *i*-th row sum of A_T ; $\sum_{i=1}^n r_i = \binom{n}{2}$.



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Lemma: For every two integers a, b satisfying $b \ge a+2 > a \ge 1$, we have

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It can be proved using $x! > (\frac{x+1}{2})^x$ for $x \ge 2$.



Proof of theorem

Observe that $\prod_{i=1}^{n} (r_i!)^{1/r_i}$ achieves the maximum when all r_i 's are almost equal. We get

$$F(T) \le (1+o(1))\frac{\sqrt{\pi}}{\sqrt{2}e}n^{3/2}\frac{(n-1)!}{2^n}$$



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Construct a new tournament T' for T by adding a new vertex v, where the edges from v to T are oriented randomly and independently. Every Hamiltonian path in T can be extended to a Hamiltonian cycle in T' with probability $\frac{1}{4}$. We have

$$P(T) \le \frac{1}{4}C(T') = O\left(n^{3/2}\frac{n!}{2^{n-1}}\right).$$



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$$I = \{ v \in V \colon vw \in E \Rightarrow \sigma(v) < \sigma(w) \}.$$

Then I is an independent set. Let X_v be the indicator random variable for $v \in I$.

$$E(X_v) = \Pr(v \in I) = \frac{1}{d_v + 1}.$$
$$\alpha(G) \ge E(|I|) = \sum_v \frac{1}{d_v + 1}.$$



Topic Course on Probabilistic Methods (week 2)



Turán Theorem



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Turán Theorem: For n = km + r ($0 \le r < k$),

$$t(n, K_{k+1}) = m^2 \binom{k}{2} + rm(k-1) + \binom{r}{2}.$$

The equality holds if and only if G is the complete k-partite graph with equitable partitions, denoted by $G_{n,k}$.







For any $k \leq n$, let q, r satisfy n = kq + r, $0 \leq r < k$. Let $e = r \binom{q+1}{2} + (m-r) \binom{q}{2}$.







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Dual version of Turán Theorem: If G has n vertices and e edges. Then $\alpha(G) \ge k$ and the equality holds if and only if $G = \overline{G}_{n,k}$.







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Proof: By Caro-Wei's theorem, $\alpha(G) \ge \sum_{v} \frac{1}{d_{v}+1}$. The minimum of $\sum_{v} \frac{1}{d_{v}+1}$ is reached as the d_{v} as close together as possible.



Topic Course on Probabilistic Methods (week 2)





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When the equality holds, I is a constant. G can not contain an induced P_2 . Therefore $G = \overline{G}_{n,k}$.







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 - Erdős-Bondy-Simonovits (1963,1974): $t(n, C_{2k}) \le ckn^{1+1/k}$.





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- Conjecture: $t(n, C_{2k}) \ge cn^{1+1/k}$ for k = 4 and $k \ge 6$.
- Conjecture (\$250 for proof and \$100 for disproof:) Suppose H is a bipartite graph. Prove or disprove that $t(n, H) = O(n^{3/2})$ if and only if H does not contain a subgraph each vertex of which has degree > 2.

