



Probabilistic Methods in Combinatorics Lecture 3

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- For $S \subset V$, let $h(S) = \sum_{F \subset S} h(F)$.
- A k -set F is crossing if it contains precisely one point from each V_i .

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

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

Here $c_k > 0$, independent of n .



A Lemma

Lemma: *Let P_k be the set of all homogeneous polynomials $f(p_1, \dots, p_k)$ of degree k with all coefficients have absolute value at most one and $p_1 p_2 \cdots p_k$ having coefficient one. Then for all $f \in P_k$ there exists $p_1, \dots, p_k \in [0, 1]$ with*

$$|f(p_1, \dots, p_k)| \geq c_k.$$

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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

$$c_k = M(f_0) > 0. \quad \square$$



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Say F has type (a_1, \dots, a_k) if $|F \cap V_i| = a_i$, $1 \leq i \leq k$. For these F ,

$$\mathbb{E}(X_F) = h(F)p_1^{a_1} \cdots p_k^{a_k}.$$



continue

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



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Let $f(p_1, \dots, p_k) = \frac{1}{n^k} E(X)$. Then $f \in P_k$.



continue

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



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Then $\mathbf{E}(|X|) \geq |\mathbf{E}(X)| \geq c_k n^k$.

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



Balancing vectors

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

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

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

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



Proof

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



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$$\begin{aligned} \mathbb{E}(X) &= \mathbb{E}\left(\sum_{i,j=1}^n \epsilon_i \epsilon_j v_i \cdot v_j\right) \\ &= \sum_{i,j=1}^n \mathbb{E}(\epsilon_i \epsilon_j) v_i \cdot v_j \\ &= \sum_{i,j=1}^n \delta_i^j v_i \cdot v_j \\ &= \sum_{i=1}^n \|v_i\|^2 = n. \end{aligned}$$



An extension

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

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



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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 $1 \leq i, j \leq n$. Then there exist $x_i, y_j = \pm 1$, $1 \leq i, j \leq n$ so that

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



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

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



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Hence,

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



Bréggman's Theorem

- $A = (a_{ij})$: an $n \times n$ matrix with all $a_{i,j} \in \{0, 1\}$.



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- r_i : the i -th row sum.

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



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- Let $A^{(1)} := A$; and $A^{(i)}$ is the submatrix obtained by deleting row $\tau(i-1)$ and column $\sigma(\tau(i-1))$ for $2 \leq i \leq n$.



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Claim: $\text{per}(A) \leq G(L)$.



continue

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 $1 \leq j \leq r$ let t_j be the permanent of A with the first row and j -th column removed (i.e., $\sigma(1) = j$). Let

$$t = \frac{t_1 + \cdots + t_r}{r} = \frac{\text{per}(A)}{r}.$$



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By induction,

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

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



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Since $\left(\prod_{j=1}^r t_j^{t_j}\right)^{\frac{1}{r}} \geq t^t$, we have

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

$$G(R_i) = (r_i!)^{1/r_i}.$$



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

