# Probabilistic Methods in Combinatorics <br> <br> Lecture 6 

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- $H$ is called strictly balanced of for any proper subgraph $H^{\prime}$,

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## Results

Theorem: Let $H$ be a balanced graph with $v$ vertices and $e$ edges. Let $A(G)$ be the event that $H$ is a subgraph (not necessarily induced) of $G$. Then $p=n^{-v / e}$ is the threshod function for $A$.

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If $p \ll n^{-v / e}$, then $\mathrm{E}(X)=o(1) ; X=0$ almost surely.
If $p \gg n^{-v / e}$, then $\mathrm{E}(X) \rightarrow \infty$. We have

$$
\Delta^{*}=O\left(\sum_{i=2}^{v} n^{v-i} p^{e-(i e / v)}\right)=o(\mathrm{E}(X))
$$

## Two other results

Theorem: Let $H$ be a strictly balanced graph with $v$ vertices and e edges and a automorphisms. Let $X$ be the copies of $H$ in $G(n, p)$. Assume $p \gg n^{-v / s}$. Then almost always

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Theorem: Let $H$ be any fixed graph. For every subgraph $H^{\prime}$ of $H$ (including $H$ itself) let $X_{H^{\prime}}$ denote the number of copies of $H^{\prime}$ in $G(n, p)$. Assume $p$ is such that $E\left(X_{H^{\prime}}\right) \rightarrow \infty$ for every $H^{\prime}$. Then almost surely

$$
X_{H} \sim \mathrm{E}\left(X_{H}\right) .
$$

## Clique number of $G(n, 1 / 2)$

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Theorem: Let $k=k(n)$ satisfying $k \sim 2 \log _{2} n$ and $f(k) \rightarrow \infty$. Then almost surely $\omega(G) \geq k$.
Proof: For each $k$-set $S$, let $X_{S}$ be the indicator random variable that $S$ is a clique and $X=\sum_{|S|=k} X_{S}$.

$$
\mathrm{E}(X)=\binom{n}{k} 2^{\binom{k}{2}}=f(k)
$$

## Continue

$$
\begin{gathered}
\Delta^{*}=\sum_{i=2}^{k-1}\binom{k}{i}\binom{n-k}{k-i} 2^{\binom{i}{2}-\binom{k}{2} .} \\
\frac{\Delta^{*}}{E(|X|)}=\sum_{i=2}^{k-1} g(i),
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where $g(i)=\frac{\binom{k}{i}\binom{n-k}{k-i}}{\binom{n}{k}} 2^{\binom{i}{2}}$. Then

$$
g(i) \leq \max \{g(2), g(k-1)\}=o\left(n^{-1}\right) .
$$

Thus, $\Delta^{*}=o(\mathrm{E}(X))$.

## Remark

$$
\frac{f(k+1)}{f(k)}=\frac{n-k}{k+1} 2^{-k} .
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For $k \sim 2 \log _{2} n$, then

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Let $k_{0}$ be the value with $f\left(k_{0}\right) \geq 1>f\left(k_{0}+1\right)$. For most of $n, f(k)$ will jump from very large to ver small. With high probabilty, $\omega(G)=k_{0}$.

## Distinct sum

A set $x_{1}, \ldots, x_{k}$ of positive integers is said to have distinct sums if all sums

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- Let $f(k)$ be the smallest $k$ for which there is a set

$$
\left\{x_{1}, x_{2}, \ldots, x_{k}\right\} \subset\{1, \ldots, n\}
$$

with distinct set.
It is clear $f(n) \geq 1+\left\lfloor\log _{2} n\right\rfloor$.

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Theorem:

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f(n)<\log _{2} n+\frac{1}{2} \log _{2} \log _{2} n+O(1) .
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## Lovász Local Lemma

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- $A_{1}, A_{2}, \ldots, A_{n}: n$ events in an arbitrary probability spaces.
- A dependency digraph $D=(V, E)$ : if for each $A_{i}, A_{i}$ is mutually independent to all the events $\left\{A_{j}: A_{i} A_{j} \notin E\right\}$.

Lovász Local Lemma, general case: If there are real number $x_{1}, \ldots, x_{n}$ such that $0 \leq x_{i}<1$ and
$\operatorname{Pr}\left(A_{i}\right) \leq x_{i} \prod_{(i, j) \in E}\left(1-x_{j}\right)$ for all $1 \leq i \leq n$. Then

$$
\operatorname{Pr}\left(\wedge_{i=1}^{n} \bar{A}_{i}\right) \geq \prod_{i=1}^{n}\left(1-x_{i}\right)>0
$$

## Symmetric Case

## Lovász Local Lemma, symmetric case: Let

$A_{1}, A_{2}, \ldots, A_{n}$ be events in an arbitrary probability space.
Suppose that each event $A_{i}$ is mutually independent of a set of all the other event $A_{j}$ but at most $d$, and that $\operatorname{Pr}\left(A_{i}\right) \leq p$ for all $1 \leq i \leq n$. If $e p(d+1)<1$, then $\operatorname{Pr}\left(\wedge_{i=1}^{n} \bar{A}_{i}\right)>0$.

## Property B

Theorem: Let $H=(V, E)$ be a hypergraph in which every edge has at least $k$ elements, and suppose that each edge of $H$ intersects at most $d$ other edges. If $e(d+1) \leq 2^{k-1}$, then $H$ has property $B$.

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Proof: Color each vertex in two colors randomly and independently. For each edge $f \in E$, let $A_{f}$ be the event that $f$ is monochromatic. Then

$$
\operatorname{Pr}\left(A_{f}\right)=2^{1-|f|} \leq 2^{1-k} .
$$

$A_{f}$ is independent to all event but at most $d$. Aplly LLL.

## $k$-coloring of $\mathbb{R}$

Let $c: \mathbb{R} \rightarrow\{1,2, \ldots, k\}$ be a $k$-coloring of $\mathbb{R}$. A set $T \subset \mathbb{R}$ is multicolored if $c(T)=\{1,2, \ldots, k\}$.

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Theorem: Let $m$ and $k$ be two positive intergers satisfying

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e(m(m-1)+1) k\left(1-\frac{1}{k}\right)^{m} \leq 1 .
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Then, for any set $S$ of $m$ real numbers there is a $k$-coloring so that each translantion $x+S$ (for $x \in \mathbb{R}$ ) is multicolored.

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The condition is satisfied if $m>(3+o(1)) k \log k$.

## Proof

First we use LLL to prove "For any finite set $X \subset \mathbb{R}$, there is a $k$-coloring so that $x+S$ (for all $x \in X$ ) is multi-colored."

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Let $Y=\cup_{x \in X}(x+S)$. Color numbers in $Y$ in $k$-colors randomly and independently. Let $A_{x}$ be the event that $x+S$ is not multi-colored.

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\operatorname{Pr}\left(A_{x}\right) \leq k\left(1-\frac{1}{k}\right)^{m-1}
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$A_{x}$ depends on $A_{y}$ if $(x+S) \cap(y+S) \neq \emptyset$. Equivalently, $y-x \in S-S$. There are at most $m(m-1)$ such events.

$$
d \leq m(m-1)
$$

## continue

Apllying LLL, we get

$$
\operatorname{Pr}\left(\wedge_{x \in X} \bar{A}_{x}\right)>0 .
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Then by Tikhonov's theorem, $[k]^{\mathbb{R}}$ is compact. For any $x \in \mathbb{R}$, let

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C_{x}=\left\{c \in[k]^{\mathbb{R}}: x+S \text { is multi-colored }\right\} .
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Now $C_{x}$ is a closed set and $\cap_{x \in X} C_{x} \neq \emptyset$ for any finite $X$. Then $\cap_{x \in \mathbb{R}} C_{x} \neq \emptyset$.

## Ramsey numbers

## Theorem (Spencer, 1975)

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Best bounds for $R(r, k)$ (for fixed $r$ and $k$ large),

$$
c\left(\frac{k}{\log k}\right)^{(r+1) / 2}<R(r, k)<(1+o(1)) \frac{k^{r-1}}{\log ^{r-2} k} .
$$

