# Topic Course on Probabilistic Methods 

(Week 3)
Alterations

Linyuan Lu<br>University of South Carolina

## 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

## Alteration

- Ramsey number $R(r, r)$
- Combinatorial geometry
- Ramsey number $R(k, r)$

■ Property B problem revisited

## Alteration method

Suppose that the "random" structure does not have all desired properties but many have a few "blemishes". With a small alteration we remove the blemishes, giving the desired structures.

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If $X<\frac{n}{2}$, then we can delete at most $\frac{n}{2}$ to destroy all monochromatic $K_{r}$. Thus, $R(r, r)>\frac{n}{2}$.
This gives $R(r, r)>(1+o(1)) \frac{1}{e} r 2^{r / 2}$. $\square$

## Combinatorial geometry

- $S$ : a set of $n$ points in the unit square $[0,1]^{2}$.
- $T(S)$ : the minimum area of a triangle whose vertices are three distinct points of $S$.

Komlós, Pintz, Szemerédi (1982): There exists a set $S$ of $n$ points in the unit square such that $T(S)=\Omega\left(\frac{\log n}{n^{2}}\right)$.

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Proof: Select $2 n$ random points uniformly and independently from $[0,1]^{2}$.

- $P, Q, R$ : three random points.
- $\mu:=\triangle P Q R$ : the area of $P Q R$.


## Proof

$$
\operatorname{Pr}(x \leq|P Q| \leq x+\Delta x) \leq \pi(x+\Delta x)^{2}-\pi x^{2} \approx 2 \pi x \Delta x
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If $\mu \leq \epsilon$, then $R$ is in the region of a rectangle of width $\frac{4 \epsilon}{x}$ and length at most $\sqrt{2}$.

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Delete one vertex from each small triangle and leave at least $n$ vertices. Now no triangle has area less that $\frac{1}{100 n^{2}}$.

## Ramsey number $R(k, t)$

## Theorem: For any $0<p<1$, we have

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R(k, t)>n-\binom{n}{k} p^{\binom{k}{2}}-\binom{n}{t}(1-p)^{\binom{t}{2}}
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Proof: Color each edge independently in red or blue; the probability of being red is $p$ while the probability of being blue is $1-p$. Let $X$ be the number of red $K_{k}$ and $Y$ be the number of blue $K_{t}$.

$$
\begin{aligned}
& \mathrm{E}(X)=\binom{n}{k} p^{\binom{k}{2}} \\
& \mathrm{E}(Y)=\binom{n}{t}(1-p)^{\left(\begin{array}{c}
\left(\begin{array}{c}
2
\end{array}\right)
\end{array}\right.}
\end{aligned}
$$

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Best lower bound: Kim (1995) and best upper bound: Shearer (1983).

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\frac{c t^{2}}{\ln t} \leq R(3, t) \leq(1+o(1)) \frac{t^{2}}{\ln t}
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Before Shearer's result, Ajtai-Komlós and Szemerédi (1980) proved $R(3, t) \leq \frac{c^{\prime} t^{2}}{\ln t}$.

## Recoloring

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Let $m(r)$ denote the minimum possible number of edges of an $r$-uniform hypergraph that does not have property $B$.

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Proof: For a fixed $r$-uniform hypergraph $H=(V, E)$ with $|E|=k 2^{r-1}$. Let $p \in[0,1]$ satisfying $k(1-p)^{r}+k^{2} p<1$.

## Coloring process

Here is a two-round coloring process.
■ First round: Color each vertex independently in red or blue with equal probability. It ends with a coloring with expected $k$ monochromatic edges. Let $U$ be the set of vertices in some monochromatic edges.

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■ First round: Color each vertex independently in red or blue with equal probability. It ends with a coloring with expected $k$ monochromatic edges. Let $U$ be the set of vertices in some monochromatic edges.
■ Second round: Consider vertices in $U$ sequentially in the (random) order of $V$. A vertex $u \in U$ is still dangerous if there is some monochromatic edge in the first coloring and for which no vertex has yet changed color.

- If $u$ is not dangerous, do nothing.
- If $u$ is still dangerous; with probability $p$, flip the color of $u$.


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\begin{gathered}
\operatorname{Pr}\left(A_{e}\right)=2^{-r}(1-p)^{r} \\
2 \sum_{e \in E(H)} \operatorname{Pr}\left(A_{e}\right)=k(1-p)^{r} .
\end{gathered}
$$

## Estimating $\operatorname{Pr}\left(C_{e}\right)$

For two edge $e, f$, we say $e$ blames $f$ if

- $e \cap f=\{v\}$ for some $v$.
- In the first coloring $f$ was blue and in the final coloring $e$ was red.
- $v$ was the last vertex of $e$ that changed color from blue to red.
- When $v$ changed its color $f$ was still entire blue.


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- When $v$ changed its color $f$ was still entire blue.

Call this event $B_{e f}$. Then

$$
\sum_{e} \operatorname{Pr}\left(C_{e}\right) \leq \sum_{e \neq f} \operatorname{Pr}\left(B_{e f}\right) .
$$

## continue

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\operatorname{Pr}\left(B_{e f} \mid \sigma\right) \leq \frac{p}{2} 2^{-r+1}(1-p)^{j} 2^{-r+1+i}\left(\frac{1+p}{2}\right)^{i}
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We have

$$
\begin{aligned}
\operatorname{Pr}\left(B_{e f}\right) & \leq 2^{1-2 r} p \mathrm{E}\left[(1+p)^{i}(1-p)^{j}\right] \\
& \leq 2^{1-2 r} p .
\end{aligned}
$$

## Estimating $k$

The failure probability is at most
$2 \sum_{e \in E(H)}\left(\operatorname{Pr}\left(A_{e}\right)+\operatorname{Pr}\left(C_{e}\right)\right) \leq k(1-p)^{r}+k^{2} p<k e^{-p r}+k^{2} p$.

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The function $f(p)=k e^{-p r}+k^{2} p$ reaches its minimum at
$p=\frac{\ln (r / k)}{r}$. The minimum value is less than 1 if

$$
k<(1+o(1)) \sqrt{\frac{2 r}{\ln r}} .
$$

## Continuous time

Spencer modified the Radhakrishnan-Srinivasan's proof slightly. To assign a random ordering of the vertex in $V$, it is sufficient to assign each vertex $v$ a birth time $x_{v} \in[0,1]$.
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The rest of proof is the same.

