

### Lecture 9: Probabilistic Method

**Theorem 1** (Markov inequality). For a nonnegative random variable and  $a > 0$ , it holds:

$$\Pr(X \geq a) \leq \frac{\mathbf{E}(X)}{a}.$$

1: Prove Markov's inequality (fine if  $X$  has finitely many values)

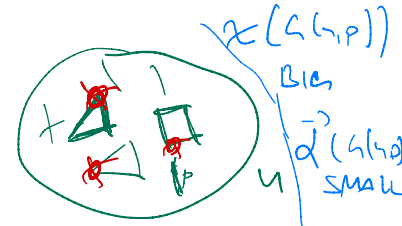
$$E(X) = \sum_i P(X=i) i \geq \sum_{i \geq a} P(X=i) a = a \cdot P(X \geq a)$$



**Theorem 2** (Erdős, 1959). For every pair of integers  $g$  and  $k$ , there exists a graph  $G$  with girth  $g(G) > g$  and chromatic number  $\chi(G) > k$ .

Let us use the following notation for the falling factorial

$$(n)_i := \binom{n}{i} i! = n(n-1)(n-2) \cdots (n-i+1).$$



*Proof.* Proof idea: Take a random graph  $\mathcal{G}(n, p)$ . Show it has few short cycles and small independent set. Small independent set means high chromatic number. Remove a vertex for each short cycle such that at least 1/2 of the vertices remain. Now no short cycles and still small independent set.

Let  $0 < \beta < 1/g$  and let  $G$  be from  $\mathcal{G}(n, p)$ , where  $p = n^{\beta-1} = \frac{1}{n^{1-\beta}}$ . Let  $X$  be the random variable that counts the number of cycles of length  $\leq g$ .

IF  $\alpha$  small  $\chi$  big  
 $\chi \geq \frac{n}{\alpha}$

2: Show that  $\mathbf{E}(X) < n/4$  for sufficiently large  $n$ .

$$E(X) \leq \sum_{i=3}^g n^i p^i = \sum_{i=3}^g n^i \cdot n^{i(\beta-1)} = \sum_{i=3}^g n^{i\beta} \leq g n^{g\beta} < \frac{n}{4}$$

$\beta < \frac{1}{g} \implies n^{g\beta} < \frac{n}{g} < \frac{n}{4}$

FOR LARGE  $n$

3: Use Markov inequality to find an upper bound probability that the number of cycles of length  $\leq g$  is at least  $n/2$ .

$$P(X \geq \frac{n}{2}) < \frac{\frac{n}{4}}{\frac{n}{2}} = \frac{1}{2}$$

Now, let  $\gamma = \lceil \frac{3}{p} \ln n \rceil$ .

4: Show that  $\Pr(\alpha(G) \geq \gamma) < \frac{1}{2}$  for sufficiently large  $n$ . Actually, it goes to zero as  $n \rightarrow \infty$ .

Hint:  $(1-p)^x \leq e^{-px}$  for any  $x > 0$ .

$$\begin{aligned} \Pr(\alpha(G) \geq \gamma) &\leq \binom{n}{\gamma} (1-p)^{\binom{\gamma}{2}} < n^\gamma e^{-p \gamma(\gamma-1)/2} \\ &= n^\gamma e^{-p \cdot (\frac{3}{p} \ln n) (\gamma-1)/2} = n^\gamma e^{-3(\gamma-1)/2} = n^{\frac{-\gamma}{2} + 3/2} \rightarrow 0 \\ &\leq \frac{1}{2} \text{ FOR LARGE } n \end{aligned}$$

Now, let  $n$  be large enough such that above two bounds are  $< 1/2$ . Then, there exists a graph  $G$  with less than  $n/2$  cycles of length  $\leq g$  and with independent set smaller than  $3n^{1-\beta} \ln n$ . From each cycle of length  $\leq g$  of  $G$  remove a vertex, this way we obtain a graph  $G'$  on at least  $n/2$  vertices, whose girth is  $> g$  and with independent number  $\alpha(G') \leq \alpha(G)$ .

5: Calculate the chromatic number of  $G'$  and finish the proof.

$$\chi(G') \geq \frac{|V(G')|}{\alpha(G')} \geq \frac{n/2}{3n^{1-\beta} \ln n} = \frac{n^\beta}{6 \ln n} \rightarrow \infty \text{ AS } n \rightarrow \infty$$

TAKE  $n$  LARGE ENOUGH TO SATISFY  $\frac{n^\beta}{6 \ln n} > K$

□

# 1 Lovász Local Lemma

Let  $A_i$  be bad events and we want to avoid all of them.

In practice the basic method is often not useful as the sum  $\sum \Pr(A_i)$  is very often bigger than 1.

In general

$$\Pr(A_1 \cup A_2 \cup \dots \cup A_n) \leq \sum_i \Pr(A_i),$$

which can easily be more than 1.

**6:** Assume that all  $A_1, \dots, A_n$  are independent and nontrivial, i.e.  $0 < \Pr(A_i) < 1$ . Improve the upper bound in the union of  $A_i$  by considering  $\bar{A}_i$ .

$$\begin{aligned} \Pr(A_1 \cup A_2 \cup \dots \cup A_n) &= 1 - \Pr(\bar{A}_1 \cap \bar{A}_2 \cap \dots \cap \bar{A}_n) \\ &= 1 - \Pr(\bar{A}_1) \cdot \Pr(\bar{A}_2) \cdot \dots \cdot \Pr(\bar{A}_n) < 1 \end{aligned}$$

$$\Pr(A \cap B) = \Pr(A) \cdot \Pr(B)$$

Lovász Local Lemma extends this when events are mostly independent.

An event  $A$  is **mutually independent** from the events  $B_1, \dots, B_k$ , if for each subset  $J \subseteq \{1, 2, \dots, k\}$ , it holds:

$$\Pr(A \cap \bigcap_{j \in J} B_j) = \Pr(A) \Pr(\bigcap_{j \in J} B_j).$$

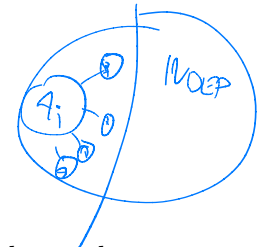
Note: It is stronger than  $A$  being independent with each  $B_i$  individually!

**Theorem 3** (Symmetric Lovász Local Lemma). Let  $A_1, A_2, \dots, A_n$  be events for which  $\Pr(A_i) \leq p$ . Suppose that each  $A_i$  mutually independent of a set of all  $A_j$ 's but except of at most other  $d$  of them. If

$$ep(d+1) \leq 1$$

then

$$\Pr\left(\bigcap_{i=1}^n \bar{A}_i\right) > 0.$$

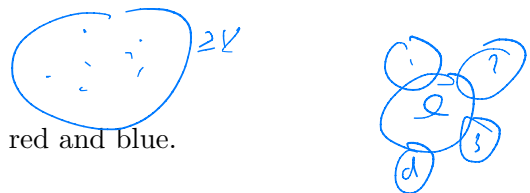


A variation of this result replaces the assumption  $ep(d+1) \leq 1$  by  $4pd \leq 1$ . Note that  $n$  does not matter in the lemma.

A coloring of a hypergraph is proper if no edge is monochromatic.



**Theorem 4.** Let  $\mathcal{H}$  be a hypergraph such that every edge contains at least  $k$  vertices and it is incident with at most  $d$  other edges. If  $e(d+1) \leq 2^{k-1}$ , then  $\mathcal{H}$  2-colorable.



*Proof.* Color the vertices of  $\mathcal{H}$  uniformly and independently by red and blue.

**7:** Finish the proof

RAVD ONLY COLOR EACH VERTEX

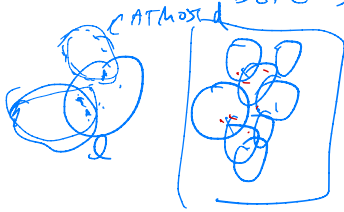
$A_e$  ... EDGE  $e$  IS MONOCHROMATIC

$$P(A_e) \leq 2 \cdot \left(\frac{1}{2}\right)^k = 2^{1-k}$$

$$e. P(d+1) \leq 1 \quad (2)$$

$A_e$  IS MUT. INDEPENDENT OF ALL BUT --  $d$  OTHER EVENTS

$$e. P(d+1) \leq 2^{k-1} \cdot 2^{1-k} = 1$$



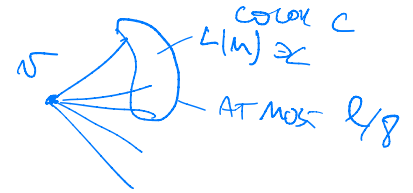
$\Rightarrow$   $\exists$  COLORING WITH NO MONOCHROMATIC EDGES  $\square$

As a particular case of the above theorem, observe that the assumption is satisfied, if every edge is incident with at most  $2^{k-3}$  others.

**Theorem 5.** Let  $G$  be a graph and  $L$  list assignment with  $|L(v)| \geq \ell > 0$  colors to every vertex  $v$ . Suppose that each color  $c$  appears in at most  $\ell/8$  lists of the neighbours of each vertex. Then,  $G$  is  $L$ -colorable.

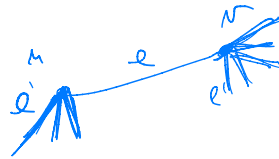
*Proof.* Color each vertex  $v$  of  $G$  independently and uniformly with a color from  $L(v)$ . Thus each of its colors is chosen with probability  $1/\ell$ . For every edge  $e = uv$  of  $G$  and every color  $c \in L(u) \cap L(v)$ , let  $A_{c,e}$  be the event that both  $u$  and  $v$  are colored by  $c$ .

8: Finish the proof.



$$P(A_{c,e}) = \frac{1}{\ell} \cdot \frac{1}{\ell} = \frac{1}{\ell^2}$$

$A_{c,e}$  ... HOW MANY "DEPENDENT" ?



$A_{c,e}$   
 $A_{c',e}$

$\ell$  CANDIDATES  $\frac{\ell}{8}$  CANDIDATES  $\frac{\ell \cdot \ell}{8} \cdot 2 = \frac{\ell^2}{4} = d$

$$4 \cdot P(d) = 4 \cdot \frac{1}{\ell^2} \cdot \frac{\ell^2}{4} = 1$$

$\Rightarrow$   $\exists$  COLORING WITH NO BAD EVENT  $\Rightarrow$  IT IS PROPER  $\square$

Let  $A_1, \dots, A_n$  be events in some probability space. A directed graph  $G$  with vertices  $\{1, \dots, n\}$  is called a *dependency graph*, if each event  $A_i$  is mutually independent of all events  $A_j$  for which there is no oriented edge  $(i, j)$  in  $G$ . Notice that the dependency graph is not necessarily uniquely determined.

**Theorem 6** (Asymmetric Lovász Local Lemma). *Let be  $A_1, \dots, A_n$  be events and  $D = (V, E)$  dependency graph of these events. For each  $i \in \{1, \dots, n\}$ , let  $x_i \in [0, 1)$  be real numbers for which*

$$\Pr(A_i) \leq x_i \prod_{ij \in E} (1 - x_j).$$

Then,

$$\Pr(\bar{A}_1 \cap \bar{A}_2 \cap \dots \cap \bar{A}_n) \geq \prod_{i=1}^n (1 - x_i) > 0.$$

**9:** Prove the symmetric version using the asymmetric version. Hint: Pick  $x_i = \frac{1}{d+1} < 1$