

## 25.1 Probabilistic Method

In this section, several theorems are listed as examples of probabilistic method.

**Theorem 25.1.1** For any  $n \in \mathbb{Z}^+$  and  $\epsilon > 0$ ,  $\exists k = \exp(\text{const.} \cdot \epsilon^2 n)$  unit vectors  $x_1, \dots, x_k$  in  $\mathbb{R}^n$  with  $|x_i \cdot x_j| \leq \epsilon$ ,  $\forall i, j \in [k]$ .

**Proof:** Assign  $+1$  or  $-1$  to each coordinate of  $x_i$  uniformly at random. Denote  $Z_l = \mathbb{1}(x_{il} = x_{jl})$  as an indicator for the  $l^{\text{th}}$  entry of vector  $x_i$  and  $x_j$  to be equal. Also we let  $Z = \sum_l Z_l$ .

Then we have

$$|x_i \cdot x_j| = \left| \sum_l Z_l - n + \sum_l Z_l \right| = 2 \left| \frac{n}{2} - Z \right|,$$

$$\mathbb{E}[Z] = \frac{n}{2}.$$

By Chernoff bound

$$\Pr[|x_i \cdot x_j| > \epsilon n] = \Pr\left[\left|\frac{n}{2} - z\right| > \frac{\epsilon}{2}n\right] \leq \exp\left(-\frac{\epsilon^2}{3} \cdot \frac{n}{2}\right).$$

Choosing  $k = \exp(\frac{\epsilon^2 n}{12})$ , we have

$$\Pr[\exists i, j \in [k] \text{ with } |x_i \cdot x_j| > \epsilon n] \leq \frac{k^2}{2} \exp\left(-\frac{\epsilon^2 n}{6}\right) = \frac{1}{2}.$$

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**Theorem 25.1.2** In any undirected graph,  $G=(V,E)$ ,  $\exists$  an independent set of size  $\geq \sum_{v \in V} \frac{1}{\text{degree}(v)+1}$ .

**Proof:** The algorithm:

- Assign independently random number drawn from  $\text{Uniform}[0, 1]$  to every vertex  $v$ .
- Output  $v$  if it is a local minimum.

Notice that the algorithm above always returns an independent set and the expected size of this independent set is  $\sum_v \frac{1}{\text{degree}(v)+1}$ .

This holds by linearity of expectation and these two observations:

1. If  $\Pr[E] > 0$ , then there exists a state of the world where  $E$  happens.

2. For any random variable  $X$ , there exists an instantiation with  $X \geq \mathbb{E}[X]$

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**Theorem 25.1.3** *In any undirected graph  $G=(V,E)$ , there exists a cut of size  $\geq \frac{|E|}{2}$*

**Proof:** The algorithm:

- Place every vertex independently in left part  $L$  with probability  $\frac{1}{2}$  and right part  $R$  with probability  $\frac{1}{2}$ .
- Return partition  $(L; R)$ .

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**Theorem 25.1.4** *For any  $k$ -CNF formula  $\phi$ , there exists an assignment that satisfies at least  $(1 - 2^{-k})$  fraction of the clauses in  $\phi$*

**Proof:** MaxSat: What is the most number of clauses we can satisfy in  $\phi$ ?

Set every variable to True or False with equal probability independently, then

$$Pr[a \text{ clause is not satisfied}] = \frac{1}{2^{\text{size of clauses}}}.$$

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## 25.2 Method of Conditional Expectation

Here is an example of satisfactory problem:

$$(x_1 \vee \neg x_2) \wedge (x_1 \vee x_2 \vee x_3 \vee x_4) \wedge (\neg x_1 \wedge \neg x_3 \vee \neg x_4).$$

Then there exists an assignment that satisfies  $\frac{3}{4} + \frac{15}{16} + \frac{7}{8}$  fraction

If we randomly assign  $x_1$  first, the expected number of fraction is

$$\begin{aligned} & Pr[x_1 = \text{True}] * \mathbb{E}[\text{fraction} | x_1 = \text{True}] + Pr[x_1 = \text{False}] * \mathbb{E}[\text{fraction} | x_1 = \text{False}] \\ &= \frac{1}{2} * \frac{11}{4} + \frac{1}{2} * \frac{19}{8}. \end{aligned}$$

We should assign  $x_1 = \text{True}$  since  $\mathbb{E}[\text{fraction} | x_1 = \text{True}]$  is greater. We can do similar analysis on  $x_2, x_3, x_4$

## 25.3 Lovasz Local Lemma

**Theorem 25.3.1** *Given a  $k$ -CNF formula  $\phi$  with number of clauses  $< 2^k$ ,  $\phi$  is always satisfied.*

**Lemma 25.3.2 (Symmetric Lovasz Local Lemma (SLL))** *Let  $E_1, E_2, \dots, E_n$  be  $n$  random events and let  $\Gamma_i$  denote the dependency neighborhood of event  $E_i$ . Suppose that  $Pr[E_i] \leq p \forall i$ , and  $|\Gamma_i| < d \forall i$  and  $e \cdot p \cdot (d + 1) \leq 1$ , where  $e$  is the base of natural logarithms. Then  $Pr[\bigcup_{i \in [n]} E_i] < 1$*

Note: The condition of this lemma is saying every event is independent of most of the events. By independency, we mean  $\forall S \subseteq [n] \setminus \{\Gamma_i \cup i\}, Pr[E_i | S] = Pr[E_i]$ .

**Theorem 25.3.3** *Let  $\phi$  be a  $k$ -CNF formula where each variable appears in at most  $\frac{2^k}{e \cdot k}$  clauses. Then  $\phi$  is satisfied.*

**Proof:** Denote  $E_i$  as clause  $i$  is not satisfied. If we set  $p = \frac{1}{2^k}$ , we have

$$d \leq k \frac{2^k}{ek} - 1 = \frac{2^k}{e} - 1.$$

Conditions for SLL are satisfied. ■