# CSCE 658: Randomized Algorithms

Lecture 14

Samson Zhou

## Probabilistic Method

- Suppose we want to argue the existence of a certain desirable object
- Existential argument, non-constructive

• If there is an algorithm that can find it, it must exist!

- What is the smallest number n = R(a, b) such that in any set of n people, there must be either:
  - *a* mutual acquaintances
  - **b** mutual strangers
- **R(a, b)** are the Ramsey numbers

 We can model a set of *n* people with a complete graph by coloring an edge (*i*, *j*) BLUE if *i* and *j* are acquaintances and GREEN if *i* and *j* are strangers

- What is the smallest number n = R(a, b) such there must be either:
  - BLUE induced complete subgraph  $K_a$
  - GREEN induced complete subgraph K<sub>b</sub>





- Finding the precise value of R(a, b) is quite difficult
- R(3,3) = 6
- R(4,4) = 18
- $\bullet 43 \leq R(5,5) \leq 48$
- $102 \le R(6,6) \le 161$
- $205 \le R(7,7) \le 497$

### Probabilistic Method for Ramsey Numbers

• If 
$$\binom{n}{k} \cdot 2^{1 - \binom{k}{2}} < 1$$
, then  $R(k, k) > n$  (Erdös)

• Consider a random coloring of  $K_n$ , so that each edge is colored BLUE with probability  $\frac{1}{2}$  and GREEN with probability  $\frac{1}{2}$ 

• For any fixed set *S* of *k* vertices, the probability *S* is monochromatic is  $\frac{1}{\binom{k}{2}} + \frac{1}{\binom{k}{2}} = 2^{1-\binom{k}{2}}$ 

## Probabilistic Method for Ramsey Numbers

- By a union bound, the probability that there exists a set of k vertices is monochromatic is  $\binom{n}{k} \cdot 2^{1-\binom{k}{2}} < 1$ .
- Then with nonzero probability, algorithm finds a coloring with no monochromatic  $K_k$
- Thus, there exists a graph coloring with no monochromatic  $K_k$
- R(k,k) > n

## Probabilistic Method

- Suppose we want to argue the existence of a certain desirable object
- Existential argument, non-constructive

• If there is an algorithm that can find it, it must exist!

## Probabilistic Method

- Suppose we want to argue the existence of a certain desirable object
- Existential argument, non-constructive

- A random variable cannot always be less than its expected value
- A random variable cannot always be more than its expected value

# Probabilistic Method for Graph Cuts

• Any undirected graph G with m edges has a cut of at least  $\frac{m}{2}$  edges

- Consider a random cut of G formed by putting each vertex into A with probability  $\frac{1}{2}$  and into B with probability  $\frac{1}{2}$
- Let the edges be e<sub>1</sub>, ..., e<sub>m</sub> and let X<sub>i</sub> denote whether e<sub>i</sub> crosses the cut

## Probabilistic Method for Graph Cuts

- The probability that  $e_i$  crosses the cut (A, B) is  $\frac{1}{2}$
- $E[X_i] = \frac{1}{2}$
- Let |C(A, B)| denote the size of the cut (A, B)
- $E[|C(A,B)|] = E[\sum_{i \in [m]} X_i] = E[X_1] + \dots + E[X_m] = \frac{m}{2}$
- Thus, there exists a cut of size  $\frac{m}{2}$

#### k-SAT

In the k-SAT problem, we are given a conjunctive normal form (CNF) formula, i.e., an AND of OR's, f(x<sub>1</sub>,...,x<sub>n</sub>) with m clauses C<sub>1</sub>,...,C<sub>m</sub> and k distinct variables per clause

• Example for k = 4:

 $(x_2 \lor \neg x_4 \lor x_5 \lor x_7) \land (x_1 \lor \neg x_3 \lor x_6 \lor x_8)$ 

• Suppose  $m < 2^k$ , we claim f must be satisfiable!

- Suppose  $m < 2^k$ , we claim f must be satisfiable!
- Suppose we assign each variable  $x_i$  a separate random TRUE/FALSE value
- For each  $i \in [m]$ , we have  $\Pr[C_i \text{ is FALSE}] \le 1/2^k$
- By a union bound

• 
$$\Pr[f(x_1, ..., x_n) = \text{FALSE}] \le \sum_{i \in [m]} \Pr[C_i \text{ is FALSE}]$$
  
 $\le \frac{m}{2^k} < 1$ 

• In the k-SAT problem, we are given a CNF formula  $f(x_1, ..., x_n)$  with m clauses  $C_1, ..., C_m$  and k distinct variables per clause

- If  $m < 2^k$ , then f is satisfiable
- What about  $m \ge 2^k$ ?

# Dependency Graph

- Let  $E_1, \ldots, E_n$  be events and let G be a graph on the vertices  $[n] \coloneqq \{1, \ldots, n\}$
- *G* is called a dependency graph for the events  $E_1, ..., E_n$  if and only if  $E_i$  is mutually independent of all events  $E_j$  for which (i, j) is not an edge in *E*

• G models the dependencies between the events  $E_1, \ldots, E_n$ 

• Theorem: Let  $E_1, \ldots, E_n$  be events and let G be their dependency graph. Suppose for all  $i \in [n]$ ,

 $\Pr[E_i] \le p, \qquad \deg(i) \le d, \qquad 4dp \le 1$ 

• Then  $\Pr[E_1^C \cap E_2^C \cap \dots \cap E_n^C] > 0$ , where  $E_i^C$  denotes the complement of  $E_i$ 

• To show  $\Pr[E_1^C \cap E_2^C \cap \dots \cap E_n^C] > 0$ , it suffices to show  $\Pr[E_i \mid E_1^C \cap E_2^C \cap \dots \cap E_{i-1}^C] \le 2p$  for all  $i \in [n]$ .

• To show  $\Pr[E_1^C \cap E_2^C \cap \dots \cap E_n^C] > 0$ , it suffices to show  $\Pr[E_i \mid E_1^C \cap E_2^C \cap \dots \cap E_{i-1}^C] \le 2p$  for all  $i \in [n]$ .

• Indeed:

 $\Pr\left[E_1^C \cap E_2^C \cap \dots \cap E_n^C\right] = \prod_{i=1}^n \Pr\left[E_i^C \mid E_1^C \cap E_2^C \cap \dots \cap E_{i-1}^C\right]$  $\geq \prod_{i=1}^n (1-2p) > 0$ 

- To show  $\Pr[E_i | E_1^C \cap E_2^C \cap \dots \cap E_{i-1}^C] \le 2p$  for all  $i \in [n]$ , we instead show  $\Pr[E_i | \bigcap_{j \in S} E_j^C] \le 2p$  for all  $|S| \le s$
- Use induction on *s*

- Our assumption is that for all  $i \in [n]$ :  $\Pr[E_i] \le p, \qquad \deg(i) \le d, \qquad 4dp \le 1$
- Base case follows from assumption for s = 1

- Assume true for s 1, show  $\Pr[E_i | \cap_{j \in S} E_j^C] \le 2p$  for all  $|S| \le s$
- Let  $\Lambda$  be the neighbors of i in G
- By joint probability,

$$\Pr[E_i \mid \bigcap_{j \in S} E_j^C] = \frac{\Pr[E_i \cap \bigcap_{j \in \Lambda} E_j^C \mid \bigcap_{j \in S \setminus \Lambda} E_j^C]}{\Pr[\bigcap_{j \in \Lambda} E_j^C \mid \bigcap_{j \in S \setminus \Lambda} E_j^C]}$$

- The numerator is  $\Pr[E_i \cap \bigcap_{j \in \Lambda} E_j^C \mid \bigcap_{j \in S \setminus \Lambda} E_j^C]$
- We have  $\Pr[E_i \cap \bigcap_{j \in \Lambda} E_j^C \mid \bigcap_{j \in S \setminus \Lambda} E_j^C] \leq \Pr[E_i \mid \bigcap_{j \in S \setminus \Lambda} E_j^C]$
- Since  $E_i$  is independent of  $E_j$  for  $j \in S \setminus \Lambda$ , then  $\Pr[E_i | \bigcap_{j \in S \setminus \Lambda} E_j^C] = \Pr[E_i] \le p$

- The denominator is  $\Pr[\bigcap_{j \in \Lambda} E_j^C | \bigcap_{j \in S \setminus \Lambda} E_j^C]$
- Our assumption is that for all  $i \in [n]$ :  $\Pr[E_i] \le p, \qquad \deg(i) \le d, \qquad 4dp \le 1$
- By a union bound,

$$\Pr\left[\bigcap_{j\in\Lambda}E_{j}^{C}\mid\bigcap_{j\in\mathcal{S}\setminus\Lambda}E_{j}^{C}\right]\geq1-\sum_{j\in\Lambda}\Pr\left[E_{j}\mid\bigcap_{j\in\mathcal{S}\setminus\Lambda}E_{j}^{C}\right]$$
$$\geq1-\sum_{j\in\Lambda}2p\geq1-2pd\geq\frac{1}{2}$$

- Assume true for s 1, show  $\Pr[E_i | \cap_{j \in S} E_j^C] \le 2p$  for all  $|S| \le s$
- Let  $\Lambda$  be the neighbors of i in G
- By conditional probability,

$$\Pr[E_i \mid \bigcap_{j \in S} E_j^C] = \frac{\Pr[E_i \cap \bigcap_{j \in \Lambda} E_j^C \mid \bigcap_{j \in S \setminus \Lambda} E_j^C]}{\Pr[\bigcap_{j \in \Lambda} E_j^C \mid \bigcap_{j \in S \setminus \Lambda} E_j^C]} \le \frac{p}{(1/2)} = 2p$$

• Theorem: Let  $E_1, \ldots, E_n$  be events and let G be their dependency graph. Suppose for all  $i \in [n]$ ,

 $\Pr[E_i] \le p, \qquad \deg(i) \le d, \qquad 4dp \le 1$ 

• Then  $\Pr[E_1^C \cap E_2^C \cap \dots \cap E_n^C] > 0$ , where  $E_i^C$  denotes the complement of  $E_i$ 

• In the k-SAT problem, we are given a CNF formula  $f(x_1, ..., x_n)$  with m clauses  $C_1, ..., C_m$  and k distinct variables per clause

- If  $m < 2^k$ , then f is satisfiable
- What about  $m \ge 2^k$ ?

# Resampling Algorithm for k-SAT

• We say clauses  $C_i$  and  $C_j$  intersect if there exists a variable  $x_k$  (or its negation) that appears in both  $C_i$  and  $C_j$ 

• Theorem: If each clause intersects with at most  $d \leq \frac{2^k}{4}$  other clauses, then f is satisfiable

# Resampling Algorithm for *k*-SAT

- Suppose we assign each variable  $x_i$  a separate random TRUE/FALSE value
- For each  $i \in [m]$ , we have  $\Pr[C_i \text{ is FALSE}] \le 1/2^k$
- If each clause intersects with at most  $d \leq \frac{2^k}{4}$  other clauses, then by the Lovász Local Lemma, the algorithm finds satisfying assignment with nonzero probability
- Thus by the probabilistic method, the assignment must be satisfiable

# Resampling Algorithm for *k*-SAT

• Suppose we assign each variable  $x_i$  a separate random TRUE/FALSE value

- As long as there is a clause  $C_j$  that is unsatisfied, we resample all the variables in  $C_j$  independently and uniformly at random
- Algorithm may never terminate?
- Algorithmic version of the Lovász Local Lemma (we will not cover this)

## Edge-Disjoint Paths

- Suppose there are *n* pairs of users who want to communicate over a network. Find a routing such that no communication paths for each pair share any edges
- Theorem: Let  $P_i$  be the set of paths that pair *i* can use. Suppose: •  $|P_i| \ge m$  for all  $i \in [n]$ 
  - For all  $i \neq j$  and any path  $P \in P_i$ , there are at most k other paths  $P' \in P_j$  that conflict with P
- If  $\frac{8nk}{m} \leq 1$ , then there exists a routing with no conflicting paths

# Edge-Disjoint Paths

- Suppose  $|P_i| = m$  and choose a random path from each  $P_i$ , independently for each  $i \in [n]$
- Let  $E_{i,j}$  be the event that the paths chosen from  $P_i$  and  $P_j$  conflict
- After fixing a path from  $P_i$ , there are at most k conflicting paths  $P_j$  among m possible paths, so that  $\Pr[E_{i,j}] \leq k/m$
- Set p = k/m in the Lovász Local Lemma

## Edge-Disjoint Paths

- Since  $E_{i,j}$  is independent of  $E_{x,y}$  for  $x, y \notin \{i, j\}$ , then each vertex in the dependency graph has degree less than 2n
- Set d < 2n in the Lovász Local Lemma
- Then  $4pd < 4\left(\frac{k}{m}\right)(2n) = \frac{8nk}{m} \le 1$
- By the Lovász Local Lemma, the algorithm finds a disjoint routing with nonzero probability
- Thus by the probabilistic method, there exists a disjoint routing