

# Classes with Randomness.

CS254

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# Outline

- Randomized Algorithms
- Randomized Complexity classes

# Randomized Algorithms

- We now begin to investigate the power of the Turing Machines which have the ability to flip coins.
- To begin we consider some randomized algorithms for some common computational problems.

# Determinants

- Recall the determinant of a matrix  $A$  is given by

$$\det A = \sum_{\pi} \sigma(\pi) \prod_{i=1}^n A_{i,\pi(i)}$$

- Here  $\pi$  is a permutation of  $1, \dots, n$  and  $\sigma(\pi)$  is 1 if  $\pi$  is a product of an even number of transpositions and -1 otherwise.

- For example, if  $A$  is the 2x2 matrix  $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$  its determinant is  $ad-bc$ .

- Determinants are useful for many things: Computing volumes, inverting matrices, etc.
- To compute a determinant of an  $n \times n$  matrix one typically uses Gaussian elimination to convert the matrix into an upper triangular form. The determinant then becomes the product of the diagonals.
- So the determinant can be computed in polynomial time.

# Determinants and Matchings

- Given a bipartite graph  $G=(V \cup U, E)$  we can compute the determinant of its adjacency matrix  $A^G$ .
- For this matrix  $a_{ij}$  entry is  $x_{ij}$  (a symbolic variable) if there is an edge from the  $i$ th element of  $V$  to the  $j$ th element of  $U$ . It is 0 otherwise.
- The only nonzero terms in the determinant correspond to perfect matchings in  $G$ .
- Since all of the elements in  $V$  and  $U$  appear at once, the terms we get in this symbolic matrix don't cancel.
- So  $G$  has a matching iff this symbolic determinant is non-zero.
- We would thus like to be able to figure out if a symbolic determinant is identically zero.
- The idea we'll use is to fill in random numbers for the variables and see if we get zero.

# Lemma

Let  $\pi(x_1, \dots, x_m)$  be a polynomial, not identically zero, in  $m$  variables each of degree at most  $d$ . Let  $M > 0$  be an integer.

Then the number of  $m$ -tuples  $(x_1, \dots, x_m) \in \{0, 1, \dots, M-1\}^m$  such that  $\pi(x_1, \dots, x_m) = 0$  is at most  $mdM^{m-1}$ .

**Proof.** By induction on  $m$ . When  $m=1$ , the lemma says that no polynomial of degree  $\leq d$  can have more than  $d$  roots which is just the Fundamental Theorem of Algebra. By induction suppose the result is true for  $m-1$  variables. Suppose  $\pi(x_1, \dots, x_m)$  is an  $m$  variable polynomial and it evaluates to 0 at some point in our domain. The highest degree coefficient of  $x_m$  is then either 0 or not. This coefficient is a polynomial in just  $x_1, \dots, x_{m-1}$ . If it is zero by our hypothesis, this can happen for at most one of  $(m-1)dM^{m-2}$  places. Since in our domain we have  $M$  choices for  $x_m$ , the total number of places where the lead term is zero is at most  $(m-1)dM^{m-1}$ . The remaining terms in  $\pi$  define a degree  $\leq d$  polynomial in  $x_m$  so can have at most  $d$  roots for each combination of  $x_1, \dots, x_{m-1}$ . This gives at most  $dM^{m-1}$  more roots. Adding these two estimates gives the result.

# A Perfect Matching Algorithm

1. Choose  $m$  random integers  $i_1, \dots, i_m$  between 0 and  $M=2m$ .
  2. Compute the determinant  $A$ ,  $\det A^G(i_1, \dots, i_m)$  by Gaussian elimination.
  3. If it is not 0 then reply  $G$  has a perfect matching
  4. Otherwise reply it does not have a perfect matching.
- Notice this algorithm might give a false negative with probability less than half.
  - By repeating the experiment multiple times we can reduce the probability to as small as we want.
  - We already had an algorithm for matching; nevertheless, the above solves the more general problem of checking when a symbolic determinant is 0, for which no deterministic  $p$ -time algorithm is known.

# Random Walks for SAT

- Randomized algorithms can also be used in the context of SAT. Consider:
  1. Start with any truth assignment  $T$ , and repeat the following  $r$  times:
    - If there is no unsatisfied clause output “Satisfiable”, halt.
    - Otherwise, take any unsatisfied clause; pick any of its literals at random and flip its value
  2. After  $r$  repetitions reply “formula is probably unsatisfiable”
    - The above algorithm is called a “random walk” algorithm -- changing a variables value can be viewed as taking a step on the boolean hypercube of truth assignments.
    - If we choose  $r$  big enough this algorithm is likely to succeed in finding a truth assignment if there is one.
    - By the coupon collector problem if  $r = 2^n n$  we can expect to have tried all possible truth assignments.
    - This is probably a big overestimate. What is a reasonable  $r$  however? To be continued next day...