Lecture 14: Schwartz-Zippel Lemma, Determinant and Randomized Complexity classes

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Schwartz-Zippel Lemma

Recall that a polynomial $p(x, y, z)$ is an expression of the form

$$14x^2y^5z^8 - 3x^3 + 17y^6z^3.$$

The degree of the polynomial is the maximum of the sums of the powers of the variables in any monomial. So in the last example, the degree is 15.

The Schwartz-Zippel Lemma turns out to be quite useful for randomized algorithms:

Lemma 1. *Let $p(x_1, \ldots, x_n)$ be a polynomial of degree d, such that p is not the 0 polynomial. Let S be any set of numbers, and let $a_1, \ldots, a_n$ be n random numbers drawn from S. Then $\Pr[p(a_1, \ldots, a_n) = 0] \leq d/|S|$.*

Proof We prove the lemma by induction on n. When $n = 1$, the theorem follows from the fact that any non-zero degree d polynomial in one variable has at most d roots. Thus $p(a) = 0$ only when a is a root, which happens with probability at most d.

For the general case. Let us write the polynomial in the form

$$p(x_1, \ldots, x_n) = x_n^\ell \cdot q(x_1, \ldots, x_{n-1}) + r(x_1, \ldots, x_n),$$

where here r is a polynomial in which the degree of x_n is at most $\ell - 1$. So we simply gather all the terms which have maximum degree in x_n.

Now let E_1 be the event that $p(a_1, \ldots, a_n) = 0$, and let E_2 be the event that $q(a_1, \ldots, a_{n-1}) = 0$. Then we have that

$$\begin{aligned}
\Pr[E_1] &= \Pr[E_1 \wedge E_2] + \Pr[E_1 \wedge \neg E_2] \\
&= \Pr[E_2] \cdot \Pr[E_2|E_1] + \Pr[\neg E_2] \cdot \Pr[E_1|\neg E_2] \\
&\leq \Pr[E_2] + \Pr[E_1|\neg E_2].
\end{aligned}$$

By induction, since q is a degree $d - \ell$ polynomial, $\Pr[E_2] \leq (d - \ell)/|S|$. Since after $x_1, \ldots, x_{n-1}$ are fixed in $\neg E_2$, we have that $p(a_1, \ldots, a_{n-1}, x_n)$ is a non-zero polynomial of degree ℓ, we have that $\Pr[E_1|\neg E_2] \leq \ell/|S|$. Thus $\Pr[E_1] \leq d/|S|$. ■

Application: Algorithm for Perfect Matching

Given a bipartite graph G with n vertices on the left and n vertices on the right, a perfect matching in the graph is a set of n disjoint edges in the graph. Here we give a simple randomized algorithm for computing whether or not a given graph contains a perfect matching.

Recall that the determinant of an $n \times n$ matrix M is defined to be

$$\mathsf{det}(M) = \sum_{\pi \in S_n} \mathsf{sign}(\pi) \prod_{i=1}^{n} M_{i\pi(i)},$$

where here S_n is the set of permutations on n elements, and $\mathsf{sign}(\pi)$ is either 1 or -1 depending on the permutation. We have algorithms for computing the determinant that run in time $O(n^3)$.

Now consider the matrix obtained from the input graph by setting

$$M_{ij} = \begin{cases} x_{ij} & \text{if } (i,j) \text{ is an edge,} \\ 0 & \text{otherwise.} \end{cases}$$

Then we have that $\mathsf{det}(M)$ is non-zero if and only if the graph has a perfect matching! Thus to test whether or not the graph has a perfect matching, it is enough to determine whether the polynomial $det(M)$ is non-zero or not. Observe that $\mathsf{det}(M)$ is a polynomial of degree at most n. Calculating this polynomial explicitly is too time consuming, since in general it may have an exponential number of monomials. Instead the following randomized algorithm works:

Input: A bipartite graph G with n vertices on each side.
Result: Whether or not G contains a perfect matching
For $i, j \in [n]$, sample a_{ij} uniformly at random from the set $\{1, 2, \ldots, 10n\}$;
Set

$$A_{ij} = \begin{cases} a_{ij} & \text{if } (i,j) \text{ is an edge,} \\ 0 & \text{otherwise ;} \end{cases}$$

if $\mathsf{det}(A) = 0$ **then**
  Output "No perfect matching";
else
  Output "There is a perfect matching";
end

Algorithm 1: Algorithm for deciding perfect matching

If the graph has no perfect matching, then clearly the polynomial $\mathsf{det}(M) = 0$, so the algorithm always outputs that there is no perfect

matching. However, when the graph does contain a perfect matching, the probability that $\det(A) = 0$ is at most $1/10$ by the Schwartz-Zippel lemma.

Randomized Classes

There are several different ways to define complexity classes involving randomness. A turing machine with access to randomness is just like a normal turing machine, except it is allowed to toss a random coin in each step, and read the value of the coin that was tossed.

BPP

We say that the randomized machine computes the function f if for every input x, $\Pr_r[M(x,r) = f(x)] \geq 2/3$, where the probability is taken over the random coin tosses of the machine M. **BPP** is the set of functions that are computable by polynomial time randomized turing machines in the above sense.

RP

We shall say that $f \in$ **RP** if there is a randomized machine that always compute the correct value when $f(x) = 0$, and computes the correct value with probability at least $2/3$ when $f(x) = 1$.

ZPP

Finally, we define the class **ZPP** to be the set of boolean functions that have an algorithm that *never* makes an error, but whose *expected* running time is polynomial in n.