# Lecture 14: The Schwartz-Zippel Lemma and the Determinant 

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## Randomness vs non-determinism

Although we cannot show that BPP $\subseteq \mathbf{N P}$, we can show that an algorithm for $3 S A T$ would give a way to simulate all randomized algorithms deterministically. This is captured by the following result:
Theorem 1. BPP $\subseteq \mathbf{N P}^{\text {SAT }}$.
Proof Suppose $f \in \mathbf{B P P}$. Let us first reduce the error of the probabilistic algorithm for $f$ to $2^{-n}$. Suppose the algorithm uses $m$ random bits. Thus, we just need to be able to distinguish the case when $M(x, r)$ accepts $1-2^{-n}$ fraction of all $m$ bit strings from the case when it accepts only $2^{-n}$ fraction of all $m$ bit strings. Distinguishing the fractions 1 from 0 would be easy (just try a single string). Distinguishing the fractions 1 from $<1$ can be done with a query to SAT. So we shall reduce to this case.

Let $u_{1}, \ldots, u_{k} \in\{0,1\}^{m}$ be $k$ random $m$ bit strings, where $k$ will be chosen to be much smaller than $2^{n}$. Then we have the following claims, where here $r \oplus u_{i}$ denotes the bitwise parity of the $m$-bit string $r$ with the $m$-bit string $u_{i}$.
Claim 2. If $f(x)=0$, for every choice of $u_{1}, \ldots, u_{k}$, there exists some $r \in\{0,1\}^{m}$ such that $\bigvee_{i} M\left(x, r \oplus u_{i}\right) \neq 1$.

The claim following from the union bound. For every choice of $u_{1}, \ldots, u_{k}$, if you pick a random $r$, the probability that $M\left(x, r \oplus u_{i}\right)$ is incorrect is at most $2^{-n}$. Thus the probability that any of them is wrong is at most $k 2^{-n}<1$.

In the other case, we prove that the opposite happens:
Claim 3. If $f(x)=1$, there exist choices $u_{1}, \ldots, u_{k}$, such that for every $r \in\{0,1\}^{m}, \vee_{i} M\left(x, r \oplus u_{i}\right)=1$.

For any fixed $r$, the probability that all choices of $u_{i}$ fail to give the correct answer is at most $2^{-n k}$. Thus, as long as $n k>m$, by the union bound some choice of $u_{i}$ will work for all choices of $r$.

Our final algorithm in $\mathbf{N P}^{\text {SAT }}$ is as follows. We start by guessing $u_{1}, \ldots, u_{k}$ (say $k=m^{2}$ )to satisfy Claim 3. Then we use the SAT oracle to check whether or not there is an $r$ that makes $M\left(x, r \oplus u_{i}\right)$ accept for some $i$.

## Schwartz-Zippel Lemma

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

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

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 4. Let $p\left(x_{1}, \ldots, x_{n}\right)$ 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 $\operatorname{Pr}\left[p\left(a_{1}, \ldots, a_{n}\right)=0\right] \leq d /|S|$.

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

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

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 $p\left(a_{1}, \ldots, a_{n}\right)=0$, and let $E_{2}$ be the event that $q\left(a_{1}, \ldots, a_{n-1}\right)=0$. Then we have that

$$
\begin{aligned}
\operatorname{Pr}\left[E_{1}\right] & =\operatorname{Pr}\left[E_{1} \wedge E_{2}\right]+\operatorname{Pr}\left[E_{1} \wedge \neg E_{2}\right] \\
& =\operatorname{Pr}\left[E_{2}\right] \cdot \operatorname{Pr}\left[E_{2} \mid E_{1}\right]+\operatorname{Pr}\left[\neg E_{2}\right] \cdot \operatorname{Pr}\left[E_{1} \mid \neg E_{2}\right] \\
& \leq \operatorname{Pr}\left[E_{2}\right]+\operatorname{Pr}\left[E_{1} \mid \neg E_{2}\right] .
\end{aligned}
$$

By induction, since $q$ is a degree $d-\ell$ polynomial, $\operatorname{Pr}\left[E_{2}\right] \leq(d-$ $\ell) /|S|$. Since after $x_{1}, \ldots, x_{n-1}$ are fixed in $\neg E_{2}$, we have that $p\left(a_{1}, \ldots, a_{n-1}, x_{n}\right)$ is a non-zero polynomial of degree $\ell$, we have that $\operatorname{Pr}\left[E_{1} \mid \neg E_{2}\right] \leq$ $\ell /|S|$. Thus $\operatorname{Pr}\left[E_{1}\right] \leq d /|S|$.

## 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

$$
\operatorname{det}(M)=\sum_{\pi \in S_{n}} \operatorname{sign}(\pi) \prod_{i=1}^{n} M_{i \pi(i)}
$$

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

Now consider the matrix obtained from the input graph by setting

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

Then we have that $\operatorname{det}(M)$ 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 $\operatorname{det}(M)$ is non-zero or not. Observe that $\operatorname{det}(M)$ 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 $a_{i j}$ uniformly at random from the set $\{1,2, \ldots, 10 n\}$;
Set

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

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    if \(\operatorname{det}(A)=0\) 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 $\operatorname{det}(M)=0$, so the algorithm always outputs that there is no perfect matching. However, when the graph does contain a perfect matching, the probability that $\operatorname{det}(A)=0$ is at most $1 / 10$ by the SchwartzZippel lemma.

