

Today we prove the Valiant-Vazirani theorem.

**Theorem 1 (Valiant-Vazirani)** *Suppose there is a polynomial time algorithm that on input a CNF formula having exactly one satisfying assignment finds that assignment. (We make no assumption on the behaviour of the algorithm on other inputs.) Then  $\mathbf{NP} = \mathbf{RP}$ .*

## 1 The Valiant-Vazirani Theorem

As discussed in the last lecture, our approach is the following: given a satisfiable formula  $\phi$  and a number  $k$  such that  $\phi$  has roughly  $2^k$  satisfying assignments, we pick a random hash function  $h : \{0, 1\}^n \rightarrow \{0, 1\}^{k+2}$  from a family of pairwise independent hash functions, and we construct a formula  $\psi(x)$  which is equivalent to  $\phi(x) \wedge (h(x) = \mathbf{0})$ . With constant probability,  $\psi$  has precisely one satisfying assignment, and so we can pass it to our hypothetical algorithm, which finds a satisfying assignment for  $\psi$  and hence a satisfying assignment for  $\phi$ .

If we are only given  $\phi$ , we can try all possible values of  $k$  between 0 and  $n$  (where  $n$  is the number of variables in  $\phi$ ), and run the above procedure for each  $k$ . When the correct value of  $k$  is chosen, we have a constant probability of finding a satisfying assignment for  $\phi$ .

Once we have a randomized algorithm that, given a satisfiable formula, finds a satisfying assignment with constant probability, we have an  $\mathbf{RP}$  algorithm for 3SAT: run the assignment-finding algorithm, accept if it finds a satisfying assignment and reject otherwise. The existence of an  $\mathbf{RP}$  algorithm for 3SAT implies that  $\mathbf{NP} \subseteq \mathbf{RP}$  because  $\mathbf{RP}$  is closed under many-to-one reductions, and so  $\mathbf{RP} = \mathbf{NP}$  because we have  $\mathbf{RP} \subseteq \mathbf{NP}$  by definition.

The main calculation that we need to perform is to show that if we have a set of size roughly  $2^k$ , and we hash its elements pairwise independently to  $\{0, 1\}^{k+2}$ , then there is a constant probability that exactly one element is hashed to  $(0, \dots, 0)$ .

**Lemma 2** *Let  $T \subseteq \{0, 1\}^n$  be a set such that  $2^k \leq |T| < 2^{k+1}$  and let  $H$  be a family of pairwise independent hash functions of the form  $h : \{0, 1\}^n \rightarrow \{0, 1\}^{k+2}$ . Then if we pick  $h$  at random from  $H$ , there is a constant probability that there is a unique element  $x \in T$  such that  $h(x) = \mathbf{0}$ . Precisely,*

$$\mathbb{P}_{h \in H} [|\{x \in T : h(x) = \mathbf{0}\}| = 1] \geq \frac{1}{8}$$

PROOF: Let us fix an element  $x \in T$ . We want to compute the probability that  $x$  is the *unique* element of  $T$  mapped into  $\mathbf{0}$  by  $h$ . Clearly,

$$\mathbb{P}_h[h(x) = \mathbf{0} \wedge \forall y \in T - \{x\}.h(y) \neq \mathbf{0}] = \mathbb{P}_h[h(x) = \mathbf{0}] \cdot \mathbb{P}_h[\forall y \in T - \{x\}.h(y) \neq \mathbf{0} | h(x) = \mathbf{0}]$$

and we know that

$$\mathbb{P}_h[h(x) = \mathbf{0}] = \frac{1}{2^{k+2}}$$

The difficult part is to estimate the other probability. First, we write

$$\mathbb{P}_h[\forall y \in T - \{x\}.h(y) \neq \mathbf{0} | h(x) = \mathbf{0}] = 1 - \mathbb{P}_h[\exists y \in T - \{x\}.h(y) = \mathbf{0} | h(x) = \mathbf{0}]$$

And then observe that

$$\begin{aligned} & \mathbb{P}_h[\exists y \in T - \{x\}.h(y) = \mathbf{0} | h(x) = \mathbf{0}] \\ & \leq \sum_{y \in |T| - \{x\}} \mathbb{P}_h[h(y) = \mathbf{0} | h(x) = \mathbf{0}] \\ & = \sum_{y \in |T| - \{x\}} \mathbb{P}_h[h(y) = \mathbf{0}] \\ & = \frac{|T| - 1}{2^{k+2}} \\ & \leq \frac{1}{2} \end{aligned}$$

Notice how we used the fact that the value of  $h(y)$  is independent of the value of  $h(x)$  when  $x \neq y$ .

Putting everything together, we have

$$\mathbb{P}_h[\forall y \in T - \{x\}.h(y) \neq \mathbf{0} | h(x) = \mathbf{0}] \geq \frac{1}{2}$$

and so

$$\mathbb{P}_h[h(x) = \mathbf{0} \wedge \forall y \in T - \{x\}.h(y) \neq \mathbf{0}] \geq \frac{1}{2^{k+3}}$$

To conclude the argument, we observe that the probability that there is a unique element of  $T$  mapped into  $\mathbf{0}$  is given by the sum over  $x \in T$  of the probability that  $x$  is the unique element mapped into  $\mathbf{0}$  (all these events are disjoint, so the probability of their union is the sum of the probabilities). The probability of a unique element mapped into  $\mathbf{0}$  is then at least  $|T|/2^{k+3} > 1/8$ .  $\square$

**Lemma 3** *There is a probabilistic polynomial time algorithm that on input a CNF formula  $\phi$  and an integer  $k$  outputs a formula  $\psi$  such that*

- *If  $\phi$  is unsatisfiable then  $\psi$  is unsatisfiable.*
- *If  $\phi$  has at least  $2^k$  and less than  $2^{k+1}$  satisfying assignments, then there is a probability at least  $1/8$  then the formula  $\psi$  has exactly one satisfying assignment.*

PROOF: Say that  $\phi$  is a formula over  $n$  variables. The algorithm picks at random vectors  $a_1, \dots, a_{k+2} \in \{0, 1\}^n$  and bits  $b_1, \dots, b_{k+2}$  and produces a formula  $\psi$  that is equivalent to the expression  $\phi(x) \wedge (a_1 \cdot x + b_1 = 0) \wedge \dots \wedge (a_{k+2} \cdot x + b_{k+2} = 0)$ . Indeed, there is no compact CNF expression to compute  $a \cdot x$  if  $a$  has a lot of ones, but we can proceed as follows: for each  $i$  we add auxiliary variables  $y_1^i, \dots, y_n^i$  and then write a CNF condition equivalent to  $(y_1^i = x_1 \wedge a_i[1]) \wedge \dots \wedge (y_n^i = y_{n-1}^i \oplus (x_n \wedge a_i[n] \oplus b_i))$ . Then  $\psi$  is the AND of the clauses in  $\phi$  plus all the above expressions for  $i = 1, 2, \dots, k+2$ .

By construction, the number of satisfying assignments of  $\psi$  is equal to the number of satisfying assignments  $x$  of  $\phi$  such that  $h_{a_1, \dots, a_{k+2}, b_1, \dots, b_{k+2}}(x) = \mathbf{0}$ . If  $\phi$  is unsatisfiable, then, for every possible choice of the  $a_i$ ,  $\psi$  is also unsatisfiable.

If  $\phi$  has between  $2^k$  and  $2^{k+1}$  assignments, then Lemma 2 implies that with probability at least  $1/8$  there is exactly one satisfying assignment for  $\psi$ .  $\square$

We can now prove the Valiant-Vazirani theorem.

PROOF:[Of Theorem 1] It is enough to show that, under the assumption of the Theorem, 3SAT has an **RP** algorithm.

On input a formula  $\phi$ , we construct formulae  $\psi_0, \dots, \psi_n$  by using the algorithm of Lemma 3 with parameters  $k = 0, \dots, n$ . We submit all formulae  $\psi_0, \dots, \psi_n$  to the algorithm in the assumption of the Theorem, and accept if the algorithm can find a satisfying assignment for at least one of the formulae. If  $\phi$  is unsatisfiable, then all the formulae are always unsatisfiable, and so the algorithm has a probability zero of accepting. If  $\phi$  is satisfiable, then for some  $k$  it has between  $2^k$  and  $2^{k+1}$  satisfying assignments, and there is a probability at least  $1/8$  that  $\psi_k$  has exactly one satisfying assignment and that the algorithm accepts. If we repeat the above procedure  $t$  times, and accept if at least one iteration accepts, then if  $\phi$  is unsatisfiable we still have probability zero of accepting, otherwise we have probability at least  $1 - (7/8)^t$  of accepting, which is more than  $1/2$  already for  $t = 6$ .  $\square$