

## #1 Verifiers and $NP$

Today we will talk about our *second* major super-duper important complexity class.

**Definition 1.** A verifier for a language  $A$  is a TM  $V$ , where

$$A = \{w \mid V \text{ accepts } \langle w, c \rangle \text{ for some string } c\}$$

We measure the time of a verifier only in terms of the length of  $w$ , so a polynomial time verifier runs in polynomial time in the length of  $w$ . A language  $A$  is polynomially verifiable if it has a polynomial time verifier.

The string  $c$  is called the certificate or proof of membership in  $A$ , and  $c$  must be polynomial in the length of  $w$ .

## #2 Problems in $NP$

Here is an example problem in  $NP$ .

$$\text{COMPOSITES} = \{x \mid x = pq, \text{ for integers } p, q, > 1\}$$

(In other words,  $x$  is not prime.) A verifier for COMPOSITES would be a TM  $V$  that takes in a pair  $\langle x, r \rangle$  where  $r$  divides  $x$ .  $V$  should then be able to determine, in time polynomial w.r.t. the length of  $x$ , if  $r$  is a divisor of  $x$ . If so,  $x \in \text{COMPOSITES}$ .

What does such a  $V$  look like? We can surely think of something—perhaps considering all multiples of  $r, 2r, 3r, \dots$  until we reach a multiple that is greater than or equal to  $x$  (we should only need to check up to  $r\sqrt{x}$ , iirc.)

## #3 NDTMs and NTIME

I omitted these definitions from Tuesday's lecture.

**Definition 2.** Let  $N$  be a deciding NDTM. The running time of  $N$  is the function  $f : \mathbb{N} \rightarrow \mathbb{N}$ , where  $f(n)$  is the maximum number of steps that  $N$  uses on any branch of its computation on any input of length  $n$ .

I will draw Sipser Figure 7.10 in class for the above.

Next, we can now define what it means to be “computable in polynomial time” on an NDTM.

$$\text{NTIME}(t) = \{L \mid L \text{ is a language decided by an } \mathcal{O}(t) \text{ time NDTM.}\}$$

Finally, we have the following important result.

**Theorem 1.**

$$NP = \bigcup_k \text{NTIME}(n^k)$$

That is, a language is in  $NP$  iff it is decided by some NDTM in polynomial time.

*Proof.* We show how to convert a poly-time verifier to an equivalent poly-time NTM and vice versa. In one direction, let  $A \in NP$  and suppose  $V$  is its verifier, running in time  $n^k$ . Construct  $N$  as follows:

$N =$  "On input  $w$  of length  $n$  :

1. nondeterministically select string  $c$  of length at most  $n^k$
2. Run  $V$  on input  $\langle w, c \rangle$
3. if  $V$  accepts, accept; otherwise, reject. "

In the other direction, suppose  $A$  is decided by a poly-time NTM and construct verifier  $V$  as follows:

$V =$  "On input  $\langle w, c \rangle$ , where  $w$  and  $c$  are strings :

1. Simulate  $N$  on  $w$ , treating each symbol of  $c$  as a description of the nondeterministic choice to make at
2. If this branch of  $N$ 's computation accepts, accept; otherwise, reject."

□

## #4 Polynomial-time reducibility

We next adapt our notions of computable functions and mapping reductions to incorporate a time component.

**Definition 3.** A function  $f : \Sigma^* \rightarrow \Sigma^*$  is computable in polynomial time if it's a computable function decided by a poly-time TM.

And next,

**Definition 4.** Language  $A$  is polynomial time reducible to language  $B$ , written  $A \leq_p B$ , if a polynomial time computable function  $f : \Sigma^* \rightarrow \Sigma^*$  is a mapping reduction from  $A$  to  $B$ .

This theorem is important.

**Theorem 2.** If  $A \leq_p B$  and  $B \in P$ , then  $A \in P$ .

I will do a proof in class if I have time.

## #5 NP-Completeness

Here is our big goal:

**Definition 5.** A language  $B$  is called NP-Complete if it satisfies two conditions:

1.  $B \in NP$ , and
2. for all  $A \in NP$ , we have  $A \leq_p B$ .

And here is the big theorem to go with NP-completeness.

**Theorem 3.** If  $B$  is NP-complete and  $B \in P$ , then  $P = NP$ .

## #6 SAT

The first problem to be shown to be NP-complete is called SAT, or the satisfiability problem. This is a very important problem. We will take it slowly.

First consider the grammar of boolean formulae, with start symbol  $B$ . Let  $\Sigma$  be an alphabet of character symbols. Let's introduce some shorthand: we write  $V \in \Sigma^*$  to mean, for each  $\sigma_i \in \Sigma$ :

$$V \rightarrow \sigma_1 V \mid \sigma_2 V \mid \dots \mid \sigma_n V \mid \sigma_1 \mid \sigma_2 \mid \dots \mid \sigma_n$$

Now, define the grammar of boolean formulae as follows.

$$\begin{aligned} V &\in \Sigma^* \\ B &\rightarrow 0 \mid 1 \mid V \mid \neg B \mid B \vee B \mid B \wedge B \end{aligned}$$

A string generated by  $B$  is called a boolean formula. The operations of disjunction ( $\vee$ ), conjunction ( $\wedge$ ), and negation ( $\neg$ ) are called boolean operations. These operations have the usual semantics.

We call a boolean formula  $\phi$  satisfiable if an assignment of 0s and 1s exist to its variables that makes the formula  $\phi$  evaluate to 1. For example:

$$\phi = (\neg x \wedge y) \vee (x \wedge \neg z)$$

$\phi$  has a satisfying assignment of  $x = 0$ ,  $y = 1$ , and  $z = 0$ .

The satisfiability problem is:

$$\text{SAT} = \{ \langle \phi \rangle \mid \phi \text{ is a satisfiable boolean formula} \}$$

Finally, here is our big theorem:

**Theorem 4.**  $\text{SAT} \in P$  iff  $P = NP$ .

And we use NP-completeness to show this fact. (This is where we're headed.)