

# *P* and *NP*

Lecture in INF4130

Department of Informatics

October 25th, 2018

# Recap from Lecture 1

- Undecidability: no Turing Machine decides  $L$
- Proving undecidability
  - First, we proved that the Halting problem is undecidable
  - Later, we proved more undecidability-results via *reductions*
- Our Turing machines had no resource-restrictions. We were interested in deciders, halting in a finite number of steps.

# Today

- Not all deciders are useful in practice. Some take too much time, or use too much memory.
- We will introduce resource-limitations.
  - Time: How many steps does a decider for  $L$  use?
  - Space: How many tape-cells does a decider for  $L$  use? (Not a part of our focus in this course.)

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- We will define *complexity classes*, classes of problems of similar difficulty.
  - P: Class of languages decidable in *polynomial time* on a *deterministic* Turing machine.
  - NP: Class of languages decidable in *polynomial time* on a *nondeterministic* Turing machine.

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  - P: Class of languages decidable in *polynomial time* on a *deterministic* Turing machine.
  - NP: Class of languages decidable in *polynomial time* on a *nondeterministic* Turing machine.
  - (P: Class of problems where solutions can be found quickly.)
  - (NP: Class of problems where solutions can be checked quickly.)

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### Definition (Nondeterministic Turing machines)

A *Nondeterministic Turing machine* (NTM) is like a deterministic Turing machine (DTM), but where the transition-function of a DTM only returned one next step, the transition function of a NTM returns a set of possible next steps. The computation of a DTM, can be viewed as a straight line of configurations following each other, ending in an accept/reject state, or continuing for ever (if the DTM loops). For a NTM, the computation will look like a tree of configurations. If any branch of the computation-tree accepts, the NTM accepts its input.

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$N$  is a NTM that decides  $HAMPATH$ .

$N =$  "On input  $\langle G, s, t \rangle$ :

- (1) Write down a list of  $|G|$  nodes from  $G$ . (Nondeterministically guess a path)
- (2) If the list contains repetitions, *reject*.
- (3) Check that the first node in the list is  $s$  and that the last is  $t$ . If not, *reject*
- (4) For each pair of nodes  $(a, b)$  in the list where  $a$  occurs right before  $b$ , check that  $(a, b)$  is an edge in  $G$ . If not, *reject*
- (5) All tests have been passes, *accept*."

# Time-complexity

## Definition (Time-complexity)

Let  $M$  be a deterministic decider (DTM that halts on all inputs). The *running time* or *time complexity* of  $M$  is the function  $f : \mathbb{N} \rightarrow \mathbb{N}$ , where  $f(n)$  is the maximum number of steps  $M$  uses on any input of length  $n$ . We say that  $M$  runs in time  $f(n)$ . We will usually use  $n$  as the length of the input,  $|w|$ .

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Notice that this gives us the *worst case* running time of  $M$ . Here  $f(n)$  equals the number of steps  $M$  uses on the "hardest" instance of length  $n$ .

Typically, we are not interested in *exact* running times. We will use *big-O notation* to abstract away minor differences in running time.

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## Example ( $L \in TIME(n^2)$ ?)

How do we show  $L \in TIME(n^2)$ ? Well, that is "easy". Just show a DTM that decides  $L$  in time  $O(n^2)$ . Okay, but how do we show that  $L \notin TIME(n^2)$ ? Must somehow show that no TM decides  $L$  quickly enough. This is challenging.

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### Definition ( $P$ (polynomial time))

Let  $P = \bigcup_{k \in \mathbb{N}} \text{TIME}(n^k)$ .  $P$  is the class of languages decidable in polynomial time on a deterministic Turing machine.

## Example (*PATH*)

Let  $PATH = \{ \langle G, s, t \rangle \mid G \text{ is a directed graph with a directed path from } s \text{ to } t \}$ .

Turns out  $PATH \in P$ .

M = "On input  $\langle G, s, t \rangle$ :

- (1) Mark node  $s$ .
- (2) Repeat until no new nodes are marked:
- (3) Look at the edges of  $G$ . If an edge  $(a, b)$  is found where  $a$  is marked and  $b$  is unmarked, mark  $b$ .
- (4) If  $t$  is marked, *accept*. If not, *reject*"

## Some problems known to be in $P$

- Given a graph, confirm that it is connected.
- Given a set of natural numbers and a goal-number  $t$ , can you pick out three elements summing to  $t$ ?
- *PRIMES* from the previous lecture (not obvious!).
- Given binary natural numbers  $a, b$  and  $c$ , does  $a * b = c$ ?

# Polynomials are robust

- $P$  is closed under complement.
- Polynomials are closed under addition.
  - $P$  is closed under intersection.
  - $P$  is closed under union.
- Polynomials are closed under multiplication.
  - $P$  is closed under concatenation.
- *All reasonable deterministic computational models are polynomially equivalent.*

## Polynomial time reductions

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Language  $A$  is *polynomial time reducible* to language  $B$ , written  $A \leq_P B$ , if there exists a polynomial time computable function  $f : \Sigma^* \rightarrow \Sigma^*$ , where for every  $w$ :  
 $w \in A \leftrightarrow f(w) \in B$ . The function  $f$  is called the polynomial (time) reduction from  $A$  to  $B$ .

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With this definition, we know that if  $A \leq_P B$ , then we can efficiently translate instances of  $A$  to instances of  $B$ .

## Theorem

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## Proof

Let  $M_B$  be the polynomial time algorithm for deciding membership in  $B$ , and let  $f$  be the polynomial reduction from  $A$  to  $B$ . We construct  $M_A$  which decides  $A$  in polynomial time as follows:

$M_A =$  "On input  $w$ :

- (1) Compute  $f(w)$ .
- (2) Run  $M_B$  on input  $f(w)$
- (3) If  $M_B$  accepts, *accept*.
- (4) If  $M_B$  rejects, *reject*."

$M_A$  gives the correct answer, since  $f$  is a reduction from  $A$  to  $B$ . Furthermore,  $M_A$  runs in polynomial time, since the all steps take polynomial time. Note that the composition of two polynomials is a polynomial. □

## Definition (Running time of NTMs)

Let  $N$  be a nondeterministic decider (NTM that halts on all inputs). The *running time* or *time complexity* of  $N$  is the function  $f : \mathbb{N} \rightarrow \mathbb{N}$ , where  $f(n)$  is the maximum number of steps  $N$  uses **on any branch of its computation** on any input of length  $n$ . We say that  $N$  runs in time  $f(n)$ .

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We will look at a second (possibly more useful) definition of  $NP$  soon, but first a familiar example.

## Example (*HAMPATH*)

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 $HAMPATH \in NP$ .

$N =$  "On input  $\langle G, s, t \rangle$ :

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This is **not** how we typically show membership in  $NP$ .

## Definition (Verifiers and Certificates)

A *verifier* for language  $L$  is an algorithm  $V$ , where  $L = \{w \mid V \text{ accepts } \langle w, c \rangle \text{ for some string } c\}$ .

We say that  $c$  is a *certificate* for membership in  $L$ . We call  $V$  a *polynomial time verifier* if it runs in polynomial time with respect to the length of  $w$ . Note that the length of  $c$  is not mentioned. So if  $V$  is a polynomial time verifier, then the certificate must be polynomial (in the length of  $w$ ) since if it was longer,  $V$  would not have time to read it.

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Why?

# Think certificates, not verifiers!

A verifier can be viewed as a simple check, that checks that  $w \in L$  given the "membership proof"  $c$ . Typically, creating  $V$  is easy if we know that we have certificates.

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Can you think of certificates for yes-instances of *HAMPATH*?

Informally:  $L \in NP$  if  $L$  has "short" certificates for the yes-instances. By short we mean polynomial in  $n$ .

## Comparing the two definitions of NP

If  $L$  has a polynomial time verifier  $V$  (running in time  $n^k$ ), then we can construct a NTM deciding  $L$  as follows:

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

- (1) Nondeterministically select a string  $c$  of length at most  $n^k$ .
- (2) Run  $V$  on input  $\langle w, c \rangle$ .
- (3) If  $V$  accepts, *accept*, otherwise, *reject*."

If we instead have a NTM  $N$  deciding  $L$ , we could make a verifier  $V$  which simulated  $N$  using the certificate to chose which branch to simulate at each split. In this case, the certificate would be the accepting branch of the computation of  $N$  on  $w$ . Since  $N$  uses polynomial time,  $c$  would also be polynomial.

## Examples of languages in $NP$

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- $CLIQUE = \{\langle G, k \rangle \mid G \text{ is an undirected graph with a clique of size } k\}$
- $PARTITION = \{\langle S \rangle \mid S \text{ is a multiset of natural numbers, such that there exists a subset } S' \text{ of } S \text{ that sums to exactly half the sum of } S\}$   
(Can  $S$  be partitioned into two parts summing to the same?)

## Theorem

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## Proof

Since every DTM can be viewed as a NTM which does not branch, every language  $L$  in  $P$ , has a NTM which decides  $L$  in polynomial time. □

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We can simulate a NTM with a DTM, but our best approach gives an exponential growth in running time. Intuition: We need to simulate each branch, but even with a branching-factor of 2, the whole tree would have size  $2^{\text{running time of the NTM}}$ .

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## **The Millennium Prize Problems (from Wikipedia)**

- Hodge conjecture
- Riemann hypothesis
- **P versus NP**
- Poincaré conjecture (**solved**)
- Yang–Mills existence and mass gap
- Navier–Stokes existence and smoothness
- Birch and Swinnerton-Dyer conjecture