

# Lecture 13 – Intro to Complexity theory

NTIN071 Automata and Grammars

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*\* Adapted from the Czech-lecture slides by Marta Vomlelová with gratitude.  
The translation, some modifications, and all errors are mine.*

## Recap of Lecture 12

- the Diagonal language  $L_D$  is not recursively enumerable
- the Universal language  $L_U$ , the Universal TM: simulate any  $M$  on any  $w$
- recursive languages are closed under complement
- Post's theorem:  $L$  recursive iff both  $L, \bar{L}$  are RE
- $L_U, \bar{L}_D$  are recursively enumerable but not recursive
- reductions between decision problems
- the Halting problem is undecidable
- (Rice's thm: nontriv. properties of programs are undecidable)
- Undecidable problems about context-free grammars
- Source of undecidability: Post's correspondence problem

## Summary of Lecture 13

- time complexity
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## CHAPTER 5: INTRO TO COMPLEXITY

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# Time complexity

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# Asymptotic notation

**Big-O notation:** Let  $f, g : \mathbb{N} \rightarrow \mathbb{R}^+$ . We say that  $f(n) \in O(g(n))$ , if there exist  $C, n_0 \in \mathbb{N}^+$  such that

$$(\forall n \geq n_0) f(n) \leq C \cdot g(n)$$

e.g.  $\limsup_{n \rightarrow \infty} \frac{f(n)}{g(n)} < \infty$ . In that case we say that  $g(n)$  is an [asymptotic] **upper bound** [up to a constant multiple] for  $f(n)$ .

**Note:** Often the imprecise term ‘upper bound’ is used; sometimes you will encounter  $f(n) = O(g(n))$ .

For example,  $f(5n^3 + 2n^2 + 22n + 6) \in O(n^3)$  with  $n_0 = 10$ ,  $C = 6$ .

**Little-o notation:**  $f(n) \in o(g(n))$ , if for all  $c > 0$  there exists  $n_0 \in \mathbb{N}^+$  so that  $(\forall n \geq n_0) f(n) < c \cdot g(n)$ , i.e.  $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$ . Then we say  $f(n)$  is [asymptotically] **dominated** by  $g(n)$ .

Analogously for  $\geq$  instead of  $\leq$ :  $\Omega, \omega$ .

# Classes of time complexity

## Definition

Let  $M$  be a Turing machine that halts on every input. The **time complexity** of  $M$  is the function  $f : \mathbb{N} \rightarrow \mathbb{N}$ , where  $f(n)$  is the maximum number of computation steps for inputs of length  $n$ .

## Definition

For  $t : \mathbb{N} \rightarrow \mathbb{R}^+$ , **TIME**( $t(n)$ ) is the class of all languages decidable by a TM of time complexity in  $O(t(n))$  (i.e., always halts and for  $|w| = n$  correctly answers in at most  $O(t(n))$  steps).

**NB:** Here we mean the standard, single-tape, deterministic TM.

## Example

**Example ( $L = \{0^i 1^j \mid i \geq 0\}$  is in  $\text{TIME}(n^2)$ )**

1. check if the input is  $0^i 1^j$ , if a 0 follows a 1, reject (time  $O(n)$ )
2. return to the beginning: hidden in the constant  $O(2n) = O(n)$
3. go through the 0s, in time  $O(n^2)$ 
  - 3.1 rewrite the next 0 to  $X$
  - 3.2 find the first 1, rewrite to  $X$
  - 3.3 return to the beginning
4. if no more 0s, check that no more 1s remain and accept (if 1 found, reject) (time  $O(n)$ )

⋮

Can we do it faster?



# Can we do it faster?

**Idea:** “compare the binary representations of  $i$  and  $j$ ”,  $\log n$  bits, for each bit need to traverse through the word

**Example** ( $L = \{0^i 1^j \mid i \geq 0\}$  is also in  $\text{TIME}(n \log n)$ )

1. check if the input is  $0^i 1^j$  and even length (time  $O(n)$ )
2. iterate while there are 0s, in time  $O(n \log n)$ 
  - 2.1 rewrite every other 0 to  $X$ , then every other 1 to  $X$
  - 2.2 check if the number of remaining 0s+1s is even, if not, reject
3. if no more 0s, check that no more 1s and accept (time  $O(n)$ )

⋮

Can we do it even faster?

Can we do it even faster? Not really.

## Theorem

*Every language decidable in time  $o(n \log n)$  [on a single-tape, deterministic TM] is regular.*

[We omit the proof.]

# Multi-tape TM

## Example (Multi-tape TM for $L = \{0^i 1^i \mid i \geq 0\}$ )

- copy 0s to Tape 2
- at first 1, switch state; erase 1 from Tape 1 & 0 from Tape 2
- accept if both tapes are erased

## Lemma

*Every multi-tape Turing Machine with time complexity  $t(n)$  is equivalent to a [single-tape] Turing Machine with time complexity  $O(t^2(n))$ .*

**Proof:** Simulation of  $n$  steps of a  $k$ -tape TM can be done in  $O(n^2)$  moves since one step takes  $4n + 2k$  moves (heads at most  $2n$  fields apart, read, write, move head marks). □

# Nondeterministic time complexity

The **time complexity** of a **nondeterministic** Turing machine that always halts is defined analogously:  $f(n)$  is the maximum number of steps in **any branch** of the computation tree.

## Definition

For  $t : \mathbb{N} \rightarrow \mathbb{R}^+$ , **NTIME**( $t(n)$ ) is the class of all languages decidable by a nondeterministic TM of time complexity in  $O(t(n))$ .

(An NTM **decides**  $L$  if halts on all inputs and recognizes  $L$ .)

## Theorem

*Any nondeterministic TM of time complexity  $t(n) \geq n$ , has a deterministic equivalent of time complexity in  $2^{O(t(n))}$ .*

## Corollary

*If  $t(n) \geq n$ , then  $\text{NTIME}(t(n)) \subseteq \text{TIME}(2^{O(t(n))})$ .*

# Proof

Recall the construction: BFS of the computation graph, keep a queue of configurations to process.

- At most  $d$  possible transitions for any  $(q, X) \in (Q \setminus F) \times \Gamma$ .
- So after  $k$  steps at most  $d^k$  configurations.
- Processing one configuration can be 'hidden' in the constant.
- Therefore the simulation is in time:

$$O(t(n)d^{t(n)}) = 2^{O(t(n))}$$

- We need to simulate multiple tapes, but:

$$(2^{O(t(n))})^2 = 2^{O(2t(n))} = 2^{O(t(n))}$$



**P vs. NP**

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# The class P

## Definition

Let **P** (also **PTIME**) be the class of all languages decidable in **polynomial time** by a [single-tape, deterministic] Turing machine:

$$P = \bigcup_k \text{TIME}(n^k)$$

- Path in a graph
- Primality of an integer (Agrawal, Kayal, Saxena 2002)
- Linear programming
- Horn-SAT

(The last two are P-complete under LOGSPACE reductions.)

## Theorem ( $CFL \subseteq P$ )

*Every context free language belongs to P.*

**Proof:** Take a ChNF grammar for  $L$ . Given input  $\omega$ , run the CYK algorithm (polynomial, in  $O(n^3)$ ). □

# The class NP: verifier-based definition

## Definition

A **verifier** for a language  $L$  is an algorithm  $V$  such that:

$$L = \{w \mid \text{there exists a finite string } c \text{ such that } V \text{ accepts } \langle w, c \rangle\}$$

Such a  $c$  is called a **certificate**. It can be over any alphabet!

Complexity of verifiers is only considered wrt. the length of  $w$ : a **polynomial verifier** must halt in time  $O(|w|^k)$  for some  $k > 0$ .

Then we can assume the certificate has polynomial length (otherwise the verifier cannot even read it).

## Definition

**NP** is the class of all languages that have a polynomial verifier.

That is, there is an algorithm that works in time polynomial in  $|w|$  and when given  $w \in L$  and a certificate  $c$  validates that  $c$  is a valid certificate for  $w \in L$ .



# Hamiltonian path

A **Hamiltonian path** in a directed graph  $G$  is a directed path  $P$  that visits each vertex of  $G$  exactly once.

$\text{HAMPATH} = \{ \langle G, s, t \rangle \mid \text{there is a Ham. path from } s \text{ to } t \text{ in } G \}$

- complexity for graphs can be measured just wrt.  $|V|$  (# edges is at most quadratic, thus polynomial)
- the **certificate** is the path (sequence of vertices)
- the algorithm verifies that the sequence is indeed a path from  $s$  to  $t$  containing each vertex exactly once; this can be easily done in polynomial time wrt.  $|V|$
- for  $\overline{\text{HAMPATH}}$  we do not know whether a polynomial verifier exists (the problem is in EXPTIME)