



# Computational Complexity. Lecture 1

P vs. NP, Deterministic  
Hierarchy Theorem

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# Today

- Define NP, state P vs. NP problem
- Search problems/decision problems
- Diagonalization
- Time hierarchy theorem (only known theorem which allows to show that certain problems are not in P)

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- <http://www.win.tue.nl/~gwoegi/P-versus-NP.htm>

# Computational problems

- In a computational problem:
  - We are given an input encoded over the alphabet  $\{0,1\}$ .
  - We want to return as output a solution satisfying some property.
  - Computational problem is then defined by the property that the output has to satisfy given the input.
- 4 natural types of problems: decision, search, optimization, counting.

# Decision problems

- In a decision problem:
  - We are given an input  $x \in \{0,1\}^*$
  - We are required to return a YES/NO answer (verify whether input satisfies property)
- E.g is an undirected graph 3 colorable?
- Specify decision problems with set of inputs  $L \subseteq \{0,1\}^*$  for which the answer is YES (language)

# Search problems

- In a search problem:
  - We are given an input  $x \in \{0,1\}^*$
  - We are required to compute some answer  $y \in \{0,1\}^*$  that is in some relation to  $x$ , if such  $y$  exists
- Search problems specified with relations  $R \subseteq \{0,1\}^* \times \{0,1\}^*$ , where  $(x,y) \in R$  iff  $y$  is an admissible answer given  $x$
- For graph 3 coloring, we would want the coloring as output if it exists (more demanding). Formally relation  $R_{3\text{COL}}$  contains pairs  $(G,c) \in$  where  $G$  is 3-colorable and  $c$  is a valid 3-coloring

# P and NP

- We study asymptotic complexity of problems
- Is there “feasible” algorithm for 3 coloring?
- “feasible algorithm” = runs in poly time
- P is class of decision problems solvable in poly time
- Easier to verify than come up with solution...

# P and NP

- P is class of decision problems solvable in poly time.
- Search problem defined by a relation  $R$  is an NP search problem if there is a poly time algorithm that given  $x$  and  $y$  decides whether  $(x,y) \in R$ , and if there is a polynomial  $p$  s.t. if  $(x,y) \in R$ , then  $|y| \leq p(|x|)$ .
- Captures the fact that relation is efficiently computable and solutions (if they exist) are short.



# P and NP

- Decision problem  $L$  is in NP if
  - (Definition 1) there is some NP relation  $R$  such that  $x \in L$  iff there is  $y$  s.t.  $(x, y) \in R$
  - (Definition 2) there is a poly time algorithm  $V(.,.)$  and a polynomial  $p$  s.t.  $x \in L$  iff there is a  $y$ ,  $|y| \leq p(x)$  s.t.  $V(x, y)$  accepts.
- NP = class of NP decision problems.

# P and NP

- **Theorem 1.** NP is the set of decision problems that are solvable in poly time by a non-deterministic Turing machine.

# P and NP

- NP as a complexity class is defined as class of decision problems: easier to develop cleaner theory, complexity of decision problems completely characterizes complexity of search problems.
- **Theorem 2.** For every NP search problem there is an NP decision problem such that if the decision problem is solvable in time  $t(n)$  then the search problem is solvable in time  $O(n^{O(1)} \cdot t(n^{O(1)}))$ . In particular,  $P=NP$  iff every NP search problem is solvable in poly time.

# P and NP

- For function  $t: \mathbb{N} \rightarrow \mathbb{N}$ , we define
  - $DTIME(t(n))$  the set of decision problems that are solvable by a deterministic Turing machine within time  $t(n)$  on inputs of length  $n$
  - $NTIME(t(n))$  the set of decision problems that are solvable by a non-deterministic Turing machine within time  $t(n)$  on inputs of length  $n$
- $\mathbf{P} = \bigcup_k DTIME(O(n^k))$
- $\mathbf{NP} = \bigcup_k NTIME(O(n^k))$

# Diagonalization

- Only known way of proving separations between complexity classes
- Similar to Cantor
- Halting Problem is undecidable

# Diagonalization

- **Definition.** (Time constructible functions) A function  $t:\mathbb{N}\rightarrow\mathbb{N}$  is time constructible if there is algorithm that, on input  $n$ , computes the value  $t(n)$  in time  $O(t(n))$
- Eg. All polynomials, all combinations of exponential, polynomial and root functions are time constructible
- Not time constructible:  $t(n) = n^2$  if the number  $n$  written in binary encodes a turing machine that halts on all inputs and  $t(n) = n^2 + 1$  otherwise

# Time hierarchy theorem

- **Theorem (simple version).** For every time-constructible function  $t(\cdot)$ , there is a language  $L$  such that every algorithm that decides  $L$  must run in time  $>t(n)$  on infinitely many inputs, but  $L \in \text{DTIME}(t^{O(1)}(n))$ .

# Time hierarchy theorem

- **Theorem (alternative formulation).** For every  $f$  time-constructible function with  $f(n)\log f(n) = o(g(n))$

$$\text{DTIME}(f(n)) \subsetneq \text{DTIME}(g(n))$$



# E and EXP

- $E = \text{DTIME}(2^{O(n)})$
- $\text{EXP} = \bigcup_k \text{DTIME}(2^{O(n^k)})$

**Corollary.**  $P \neq E$  and  $E \neq \text{EXP}$

# Ladner's theorem and NP-intermediate problems

- **Theorem.** If  $P \neq NP$  then there exists a language  $L \in NP \setminus P$  that is not NP-complete.