

# Principles of Knowledge Representation and Reasoning

## Complexity Theory

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## Motivation

### Reminder: Basic Notions

- Algorithms and Turing Machines
- Problems, Solutions, and Complexity
- Complexity Classes P and NP
- Upper and Lower Bounds
- Polynomial Reductions
- NP-Completeness

### Beyond NP

- The Class co-NP
- The Class PSPACE
- Other Classes

### Oracle TMs and the Polynomial Hierarchy

- Oracle Turing-Machines
- Turing Reduction
- Complexity Classes Based on OTMs

QBE

# Motivation for Using Complexity Theory

- ▶ Complexity theory can answer questions on how easy or hard a problem is
- ▶ Gives hints on what algorithms could be appropriate, e.g.:
  - ▶ algorithms for **polynomial-time problems** are usually easy to design
  - ▶ for **NP-complete** problems, backtracking and local search work well
- ▶ Gives hints on what type of algorithm will (most probably) not work
  - ▶ for problems that are believed to be harder than NP-complete ones, simple backtracking will not work
- ▶ Gives hint on what sub-problems might be interesting

# Algorithms and Turing Machines

- ▶ We use **Turing machines** as formal models of algorithms
- ▶ This is justified, because:
  - ▶ we assume that Turing machines can compute all computable functions
  - ▶ the resource requirements (in term of time and memory) of a Turing machine are only polynomially worse than other models
- ▶ The regular type of Turing machine is the **deterministic** one: **DTM** (or simply **TM**)
- ▶ Often, however, we use the notion of **nondeterministic** TMs: **NDTM**

# Problems, Solutions, and Complexity

- ▶ A **problem** is a set of pairs  $(I, A)$  of strings in  $\{0, 1\}^*$ .  
 $I$ : Instance;  $A$ : Answer.  
If  $A \in \{0, 1\}$ : **decision problem**
- ▶ A **decision problem** is the same as a **formal language**: namely the set of strings formed by the instances with answer 1
- ▶ An algorithm **decides** (or **solves**) a problem if it computes the right answer for all instances.
- ▶ The **complexity of an algorithm** is a function

$$T: \mathbf{N} \rightarrow \mathbf{N},$$

measuring the **number of basic steps** (or memory requirement) the algorithm needs to compute an answer depending on the **size** of the instance.

- ▶ The **complexity of a problem** is the complexity of the most efficient algorithm that solves this problem.

# Complexity Classes P and NP

Problems are categorized into **complexity classes** according to the requirements of computational resources:

- ▶ The class of problems decidable on **deterministic Turing machines** in **polynomial time**: **P**
- ▶ Problems in P are assumed to be **efficiently solvable** (although this might not be true if the exponent is very large)
- ▶ In practice, this notion appears to be more often reasonable than not
- ▶ The class of problems decidable on **non-deterministic Turing machines** in **polynomial time**: **NP**
- ▶ More classes are definable using other resource bounds on time and memory

# Upper and Lower Bounds

- ▶ **Upper bounds** (**membership** in a class) are usually easy to prove:
  - ▶ provide an **algorithm**
  - ▶ show that the resource bounds are respected
- ▶ **Lower bounds** (**hardness** for a class) are usually difficult to show:
  - ▶ the technical tool here is the **polynomial reduction** (or any other appropriate reduction)
  - ▶ show that some hard problem can be reduced to the problem at hand

# Polynomial Reductions

- ▶ Given two languages  $L_1$  and  $L_2$ ,  $L_1$  can be **polynomially reduced to**  $L_2$ , written  $L_1 \leq_p L_2$ , iff there exists a polynomially computable function  $f$  such that

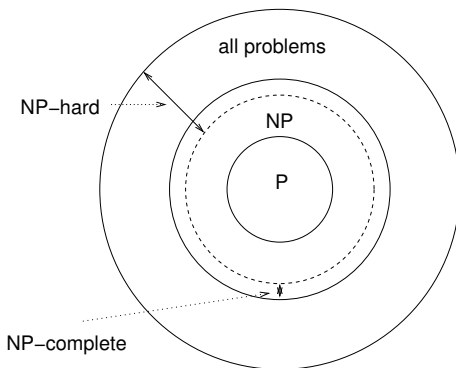
$$x \in L_1 \text{ iff } f(x) \in L_2$$

- ▶ It cannot be harder to decide  $L_1$  than  $L_2$
- ▶  $L$  is **hard** for a class  $C$  ( **$C$ -hard**) iff all languages of this class can be reduced to  $L$ .
- ▶  $L$  is **complete** for  $C$  ( **$C$ -complete**) iff  $L$  is  $C$ -hard and  $L \in C$ .



# NP-complete Problems

- ▶ A problem is **NP-complete** iff it is **NP-hard** and **in NP**.
- ▶ Example: **SAT** – the satisfiability problem for propositional logic – is NP-complete (Cook/Karp)
- ▶ Membership is obvious, hardness follows because computations on a NDTM correspond to satisfying truth-assignments of certain formulae



# The Complexity Class co-NP

- ▶ Note that there is some **asymmetry** in the definition of NP:
  - ▶ It is clear that we can decide **SAT** by using a **NDTM** with polynomially bounded computation
  - ▶ There exists an accepting computation of polynomial length iff the formula is satisfiable
  - ▶ What if we want to solve UNSAT, the complementary problem?
  - ▶ It seems necessary to check **all** possible truth-assignments!
- ▶ Define **co-C** =  $\{L \mid \Sigma^* - L \in C\}$ , provided  $\Sigma$  is our alphabet
- ▶ **co-NP** =  $\{L \mid \Sigma^* - L \in \text{NP}\}$
- ▶ For example UNSAT, TAUT  $\in$  co-NP!
- ▶ **Note:** P is closed under complement, i.e.,

$$P \subseteq \text{NP} \cap \text{co-NP}$$

# PSPACE

There are problems even more difficult than NP and co-NP.

## Definition ((N)PSPACE)

**PSPACE** (**NPSPACE**) is the class of decision problems that can be decided on deterministic (non-deterministic) Turing machines using only **polynomially many tape cells**.

Some facts about PSPACE:

- ▶ PSPACE is **closed under complements** (as all other deterministic classes)
- ▶ PSPACE is **identical** to NPSPACE (because non-deterministic Turing machines can be simulated on deterministic TMs using only quadratic space)
- ▶  $NP \subseteq PSPACE$  (because in polynomial time one can “visit” only polynomial space, i.e.,  $NP \subseteq NPSPACE$ )
- ▶ It is **unknown** whether  $NP \neq PSPACE$ , but it is **believed** that this is true.

# PSPACE-completeness

## Definition (PSPACE-completeness)

A decision problem (or language) is **PSPACE-complete**, if it is in PSPACE and all other problems in PSPACE can be polynomially reduced to it.

Intuitively, **PSPACE-complete** problems are the “hardest” problems in PSPACE (similar to NP-completeness). They appear to be “harder” than **NP-complete** problems from a *practical point of view*.

An example for a PSPACE-complete problem is the **NFA equivalence problem**:

**Instance:** *Two non-deterministic finite state automata  $A_1$  and  $A_2$ .*

**Question:** *Are the languages accepted by  $A_1$  and  $A_2$  identical?*

## Other Complexity Classes . . .

- ▶ There are complexity classes **above PSPACE** (EXPTIME, EXPSPACE, NEXPTIME, DEXPTIME . . .)
- ▶ there are (infinitely many) classes **between NP and PSPACE** (the **polynomial hierarchy** defined by **oracle machines**)
- ▶ there are (infinitely many) classes **inside P** (circuit classes with different depths)
- ▶ and for most of the classes **we do not know** whether the containment relationships are **strict**

# Oracle Turing Machines

- ▶ An **Oracle Turing machine** ((N)OTM) is a Turing machine (DTM, NDTM) with the possibility to query an **oracle** (i. e., a different Turing machine **without resource restrictions**) whether it accepts or rejects a given string.
- ▶ **Computation by the oracle does not cost anything!**
- ▶ **Formalization:**
  - ▶ a tape onto which strings for the oracle are written,
  - ▶ a yes/no answer from the oracle depending on whether it accepts or rejects the input string.
- ▶ Usage of OTMs answers **what-if questions**: What if we could solve the oracle-problem efficiently?

# Turing Reductions

- ▶ **OTMs** allow us to define a more general type of reduction
- ▶ **Idea**: The “classical” reduction can be seen as calling a subroutine once.
- ▶  $L_1$  is **Turing-reducible** to  $L_2$ , symbolically  $L_1 \leq_T L_2$ , if there exists a poly-time OTM that decides  $L_1$  by using an oracle for  $L_2$ .
- ▶ Polynomial reducibility implies Turing reducibility, but not *vice versa*!
- ▶ NP-hardness and co-NP-hardness with respect to Turing reducibility are **equivalent**!
- ▶ Turing reducibility can also be applied to general search problems!

# Complexity Classes Based on Oracle TMs

1.  $P^{NP}$  = decision problems solved by poly-time DTMs with an oracle for a decision problem in NP.
2.  $NP^{NP}$  = decision problems solved by poly-time NDTMs with an oracle for a decision problem in NP.
3.  $co-NP^{NP}$  = complements of decision problems solved by poly-time NDTMs with an oracle for a decision problem in NP.
4.  $NP^{NP^{NP}}$  = ...

... and so on



## Example

- ▶ Consider the **Minimum Equivalent Expression (MEE)** problem:

**Instance:** *A well-formed Boolean formula  $\phi$  using the standard connectives (not  $\leftrightarrow$ ) and a nonnegative integer  $K$ .*

**Question:** *Is there a well-formed Boolean formula  $\phi'$  that contains  $K$  or fewer literal occurrences and that is logical equivalent to  $\phi$ ?*

- ▶ This problem is NP-hard (wrt. to Turing reductions).
- ▶ It does not appear to be NP-complete
- ▶ We could guess a formula and then use a SAT-oracle
- ▶  $\text{MEE} \in \text{NP}^{\text{NP}}$ .

# The Polynomial Hierarchy

The complexity classes based on OTMs form an infinite hierarchy.

## The polynomial hierarchy PH

$$\begin{array}{lll} \Sigma_0^P = P & \Pi_0^P = P & \Delta_0^P = P \\ \Sigma_{i+1}^P = \text{NP}^{\Sigma_i^P} & \Pi_{i+1}^P = \text{co-}\Sigma_{i+1}^P & \Delta_{i+1}^P = P^{\Sigma_i^P} \end{array}$$

- ▶  $\text{PH} = \bigcup_{i \geq 0} (\Sigma_i^P \cup \Pi_i^P \cup \Delta_i^P) \subseteq \text{PSPACE}$
- ▶  $\text{NP} = \Sigma_1^P$
- ▶  $\text{co-NP} = \Pi_1^P$

## Quantified Boolean Formulae: Definition

- ▶ If  $\phi$  is a propositional formula,  $P$  is the set of Boolean variables used in  $\phi$  and  $\sigma$  is a sequence of  $\exists p$  and  $\forall p$ , one for every  $p \in P$ , then  $\sigma\phi$  is a **QBF**.
- ▶ A formula  $\exists x\phi$  is **true** if and only if  $\phi[\top/x] \vee \phi[\perp/x]$  is true. (Equivalently,  $\phi[\top/x]$  is true **or**  $\phi[\perp/x]$  is true.)
- ▶ A formula  $\forall x\phi$  is **true** if and only if  $\phi[\top/x] \wedge \phi[\perp/x]$  is true. (Equivalently,  $\phi[\top/x]$  is true **and**  $\phi[\perp/x]$  is true.)
- ▶ This definition directly leads to an AND/OR tree traversal algorithm for evaluating QBF.

## Quantified Boolean Formulae: Definition

The **evaluation problem of QBF** generalizes both the *satisfiability* and *validity/tautology problems* of propositional logic.

The latter are respectively **NP-complete** and **co-NP-complete** whereas the former is **PSPACE-complete**.

### Example

The formulae  $\forall x \exists y (x \leftrightarrow y)$  and  $\exists x \exists y (x \wedge y)$  are true.

### Example

The formulae  $\exists x \forall y (x \leftrightarrow y)$  and  $\forall x \forall y (x \vee y)$  are false.

# The Polynomial Hierarchy: Connection to QBF

Truth of QBFs with prefix  $\overbrace{\forall \exists \forall \dots}^i$  is  $\Pi_i^P$ -complete.

Truth of QBFs with prefix  $\overbrace{\exists \forall \exists \dots}^i$  is  $\Sigma_i^P$ -complete.

Special cases corresponding to **SAT** and **TAUT**:

The truth of QBFs with prefix  $\exists x_1^1 \dots x_n^1$  is  $\text{NP} = \Sigma_1^P$ -complete.

The truth of QBFs with prefix  $\forall x_1^1 \dots x_n^1$  is  $\text{co-NP} = \Pi_1^P$ -complete.

# Literature



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