

Complexity Theory

Knowledge Representation and Reasoning

November 2, 2005

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Motivation for Using Complexity Theory

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

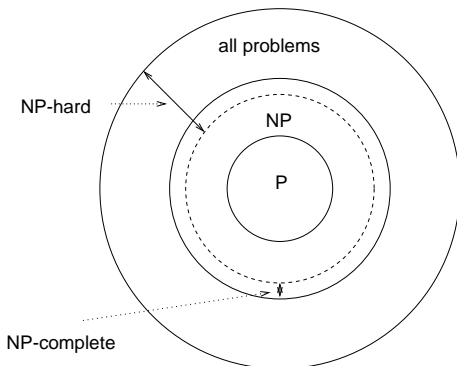
- ▶ Given two languages L_1 and L_2 , L_1 can be *polynomially reduced to* L_2 , written $L_1 \leq_p L_2$, iff there exists a polynomially computable function f such that

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

- 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 reduce to it.
- L is **complete** for C (C -complete) iff it is 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 $\text{co-}C = \{L \mid \Sigma^* - L \in C\}$, provided Σ is our alphabet
- ↪ $\text{co-NP} = \{L \mid \Sigma^* - L \in \text{NP}\}$
- ▶ For example UNSAT, TAUT \in co-NP!
- ▶ Note: P is closed under complement, i.e.,

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

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 **polynomial 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**, another Turing machine **without resource restrictions**, whether it accepts or reject 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-reduced** to L_2 , symbolically $L_1 \leq_T L_2$ if there exists an OTM that decides L_1 by using an oracle for L_2 .
- ▶ Polynomial reducibility implies Turing reducibility, but not *vice versa*!
- ↪ NP-completeness and co-NP-completeness with respect to Turing reducibility are **identical**!
- 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 ϕ using the standard connectives (not \leftrightarrow) and a nonnegative integer K .

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

- ▶ This problem is NP-hard (writ. to Turing reductions).
- ▶ It does not appear to be NP-complete
- ▶ We could guess a formula and then use a SAT-oracle

\rightsquigarrow MME \in NP^{NP}.

The Polynomial Hierarchy

The complexity classes based on OTMs form an infinite hierarchy.

The polynomial hierarchy PH

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

- ▶ **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 ϕ is a propositional formula, P is the set of Boolean variables used in ϕ and σ 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 Π_i^P -complete.

Truth of QBFs with prefix $\overbrace{\exists \forall \exists \dots}^i$ is Σ_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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