

CS:4980

Foundations of Embedded Systems

Safety Requirements

Part II

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A Brief Detour into Computational Complexity

Goal: Classify computational problems in terms of (roughly) how many basic operations it takes to solve the problem, as function of input size

Example 1: Finding maximum of a list of numbers

- Time complexity is linear: $O(n)$

Example 2: Sorting a list of numbers

- Algorithm (e.g. selection-sort) with doubly-nested loop: $O(n^2)$
- More efficient algorithm (e.g. quicksort) possible: $O(n \log n)$

A Brief Detour into Computational Complexity

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Example 3: *Expression evaluation*. Given

1. an expression e (with not/or/ and as operations) over Boolean vars, and
 2. an assignment a of 0/1 values to vars,
- determine whether e evaluates to 1 or 0.

Linear-time $O(n)$

Example 4: *Boolean satisfiability*. Given an expression e , determine if there is an assignment a to vars that makes the expression evaluate to 1

- Naïve algorithm: Evaluate e on every possible assignment a
- Exponentially many choices for a : algorithm is $O(2^k)$, k = no. of vars

The Class P

- ❑ *Polynomial-time* algorithm means an algorithm with time complexity such as $O(n)$, $O(n \log n)$, $O(n^2)$, $O(n^3)$, or $O(n^c)$, for constant c
- ❑ A problem is in **P** if there is a polynomial-time algorithm to solve it
- ❑ **Examples:**
 - Finding maximum
 - Sorting
 - Expression evaluation
 - Finding shortest path in a graph
- ❑ **P** is the class of *tractable* (i.e., efficiently solvable) problems
 - Problem can be solved exactly
 - Solution will scale reasonably well as input size grows
 - In principle, $O(n)$ is better than $O(n^2)$

NP-Complete Problems

- ❑ SAT: Given an expression e over Boolean variables, check if there exists an assignment of 0/1 values to vars for which e evaluates to 1
 - No proof that SAT is in P (no known polynomial-time algorithm)
 - No proof that SAT is not in P
- ❑ Cook (1972): SAT is *NP-complete*
- ❑ Hundreds of problems equivalent to SAT
 - Hamiltonian Path: Is there a path in a graph from source to destination that visits each vertex exactly once
 - Max Clique: Given a graph, find largest subset of vertices such that there is an edge between every pair of vertices in this set
- ❑ Grand Challenge Open Problem : Is $P = NP$?
 - If you find a polynomial-time algorithm for SAT, then $P = NP$, and many other problems will have polynomial-time algorithms
 - If you prove SAT is not in P , then $P \neq NP$, and many other problems then provably don't have efficient algorithms

NP-Completeness Continued

- Known algorithms for SAT are exponential-time in the worst-case, but
 - Highly efficient implementations, SAT solvers, exist
 - Can handle millions of variables
 - Many practical problems solved by encoding into SAT
- Key feature of NP problems such as SAT: suffices to find one satisfying assignment
- This does not hold for all intractable problems
 - Validity: Given a Boolean expression e , is it the case that e evaluates to 1 no matter what values we give to its variables
- Many complexity classes beyond NP: coNP , PSPACE , Exptime , ...
 - Problems may require exponential-time (or more) to solve
 - Not all exponential-time problems are equal.

(Un)Decidability

- ❑ Some problems cannot be solved by a computer at all!
- ❑ Fundamental Theorem of CS (Alan Turing, 1936):
 - The *Halting problem* for Turing machines is **undecidable**
There is no program that takes as its input an arbitrary program C and an arbitrary input x , and determines if C terminates on x
- ❑ **Intuition:** If a program could analyze other programs exactly, then it can analyze itself, and this suffices to set up a logical contradiction!
- ❑ A surprisingly undecidable problem: Does a given a polynomial (e.g., $x^3 + 2xy^2 - 15xy + 156$) have integer roots?
- ❑ **Decidable Problems:** There exists a program (or Turing machine) that solves the problem correctly (gives the right answer and stops)
 - Includes problems in P as well as intractable classes such as NP , $Exptime$, etc.

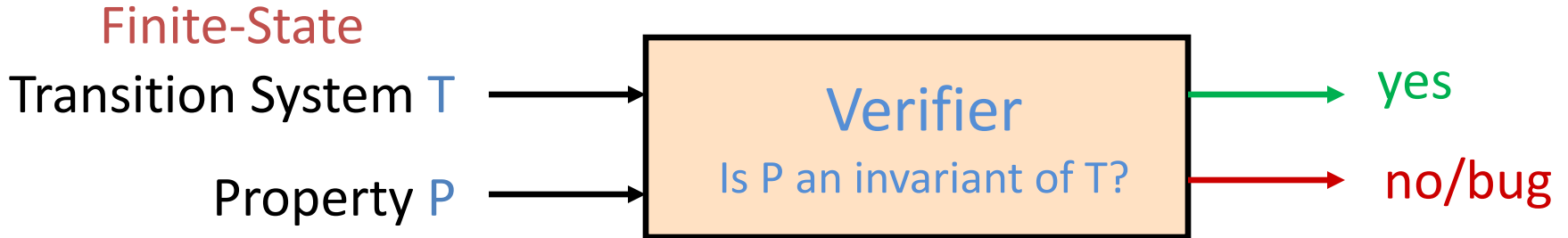
Back To Invariant Verification Problem



Theorem: The invariant verification problem is undecidable.

Proof idea: undecidable problems for Turing machines can be recast as invariant verification problems for transition systems with integer state variables

Finite-State Invariant Verification Problem

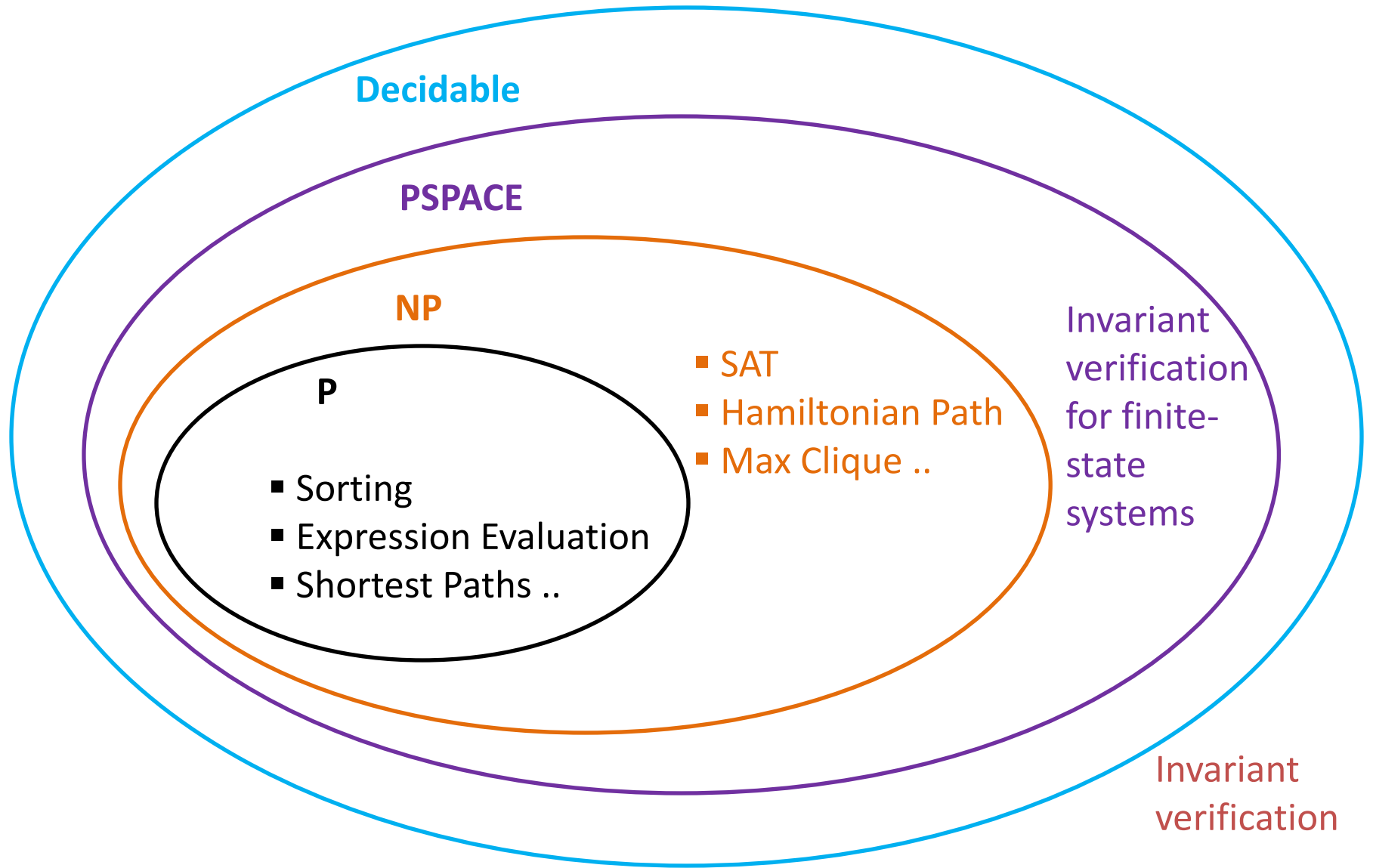


Theorem: The invariant verification problem for finite-state systems is decidable

Proof sketch: If T has k Boolean state vars, then total number of states is 2^k .

Verifier can systematically search through all possible states.

Complexity is exponential. More precisely, it is **PSPACE**, a class of problems harder than **NP-complete** problems such as SAT.



Credits

Notes based on Chapter 3 of

Principles of Cyber-Physical Systems

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