CS 420: Advanced Algorithm Design and Analysis Spring 2015 – Lecture 19

Department of Computer Science University of British Columbia



March 17, 2015

Assignments...

Asst6/7...(due March 19)

Midterm III...

- Q/A session...March 24; 5:30-7:00; DMPT 110
- Exam...March 25; 5:30-7:00; DMPT 110
- ...on all course material up to and including March 19 lecture

Announcements (cont.)

Readings...

- matchings and network flows [Kleinberg&Tardos, Chapt. 7], [Cormen et al., Chapt. 26], [Dasgupta et al., Chapter 7]
- reductions and NP-hardness [Kleinberg&Tardos, Chapt. 8, 11], [Cormen et al., Chapt. 34,35]

Last classes...

Matchings and Network Flows

- network flows
 - definitions
 - relationship with bipartite matchings
 - duality

Reductions and relative hardness of problems

- reductions...treated more formally
- overview of problems with efficient algorithms
 ... and related problems with no known efficient algorithm
- the complexity classes P and NP
- NP-hardness and NP-completeness

Reductions and relative hardness of problems

- some examples of reductions establishing NP-hardness and NP-completeness
 - ► HC S_PTSP
 - Clique $\bigcirc_P LargestCommonSubgraph$
 - ▶ VC \bigcirc_P DominatingSet
 - ► 3-SAT SPVC

left column

-spanning trees min cost maximum width

-path problems min-cost min colour-transitions Eulerian path

-graph colouring 2-colouring (bipartite) 4-colouring (planar graph)

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- -spanning trees min-cost maximum width
- -path problems min-cost min colour-transitions Eulerian path

-graph colouring 2-colouring (bipartite) 4-colouring (planar graph)

right column

bounded-degree MST bounded-diameter MST

longest (simple) path min total colours Hamiltonian path

3-colourability 3-colouring (planar graph)

left column

-matchings etc. max size (bipartite) vertex cover (bipartite) general (non-bipartite) *b*-matchings

-network flows etc.

max value integral/general capacities vertex capacities minimum cut edge/vertex-disjoint paths

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right column

3-d matching (triangle cover) maximum independent set vertex cover (tripartite)

flows with edge costs undirected flows with lower bounds vertex-disjoint connecting paths

All of the **left column** problems have *efficient* solutions: their decision versions belong to the complexity class **P**, defined to be the family of decision problems (languages) that can be decided (recognized) in time bounded by some polynomial in the input size.

Why are we interested in polynomial time?

- generous definition of tractable
- often equates to tractable in practice
- closure properties (composition)
- invariance under natural computation models

None of the **right column** problems are known to have *efficient* solutions.

Nevertheless, their decision versions all admit efficient *certification*; i.e. a short proof/certificate that the answer is YES. They all belong to the complexity class **NP** is defined to be the family of decision problems (languages) whose membership can be certified/verified in time bounded by some polynomial in the input size.

NP stands for *non-deterministic* polynomial-time: certification corresponds to acceptance by a non-deterministic machine.

Deterministic language acceptance



Machine M_L accepts L if: $\alpha \in L$ if and only if M_L outputs YES on input α

Non-deterministic language acceptance



Machine M_L non-deterministically accepts L if: $\alpha \in L$ if and only if there exists a string β such that M_L outputs YES on input (α, β) .

The complexity classes ${\bf P}$ and ${\bf NP}$

P denotes the set of languages that can be (deterministically) accepted in time bounded by some polynomial in the input length.
NP denotes the set of languages that can be (non-deterministically) accepted in time bounded by some polynomial in the input length.

It turns out that all of the **right column** problems are as hard as any problem in **NP**, up to polynomial factors, which is abbreviated **NP**-hard. Since they are also in **NP** they belong to the class **NP**-complete.

NP-complete problems have the property that they have polynomial-time solutions (i.e. they belong to **P**) if and only if **P=NP**, i.e. all problems in **NP** have polynomial-time solutions. How could we possibly show that some problem X is NP-hard? We don't even know all of the problems in NP!

- it is straightforward once we know some NP-hard problem A: simply demonstrate A S_{t(n)}X, where t(n) is some polynomial in n. (Hereafter, we write A S_PX)
- the real breakthrough was the demonstration of a *first* NP-hard problem



Reductions and relative hardness of problems

- The Cook-Levin theorem: establishing the first NP-hard problem
- more examples of reductions establishing NP-hardness and NP-completeness

Boolean Expressions

A Boolean expression over the set of Boolean variables $\{x_1, x_2, \ldots, x_n\}$ is defined (recursively) as:

- 1. a variable
- 2. the negation of a Boolean expression
- 3. the disjunction (or) of two Boolean expressions
- 4. the conjunction (and) of two logical expressions

A Boolean expression is *satisfiable* if there is an assignment of truth values to its variables such the the expression *evaluates* to true.

Conjunctive Normal Form

By applying DeMorgan's distributive laws, any Boolean expression can be converted to an equivalent expression E in *conjunctive normal form* (CNF):

$$E \equiv D_1 \wedge D_2 \wedge \ldots \wedge D_t$$

a conjunction of *disjuncts*, where each disjunct D_i has the form

$$L_{i,1} \vee L_{i,2} \vee \ldots \vee L_{i,s}$$

a disjunction of *literals*, where each literal $L_{i,j}$ is either a variable x_j or the negation of a variable \overline{x}_i .

Conjunctive Normal Form

A formula in k-CNF has the property that each of its disjuncts has at most k literals. For example:

 $(x_1 \lor \overline{x}_2) \land (\overline{x}_1 \lor \overline{x}_3) \land (x_1 \lor x_2) \land (\overline{x}_2 \lor x_4) \land (\overline{x}_1 \lor x_4)$

is a Boolean expression in 2-CNF.

Satisfiability

The language SAT is defined as the set of all satisfiable Boolean expressions. Its restriction k-SAT is the set of all satisfiable Boolean expressions in k-CNF. Note:

- \blacktriangleright 2-SAT is in P , since 2-SAT $\textcircled{}_P\text{digraph}_\text{connectivity}$
- ► k-SAT \bigcirc_P SAT and SAT \bigcirc_P 3-SAT
- ► SAT is in **NP**

Cook's Theorem...

Theorem: SAT is also NP-hard

Literally ... for every language L in NP, $L \bigotimes_P SAT$

How could this be proved?

- A language L is in NP iff there is a non-deterministic machine M that accepts strings α ∈ L in |α|^k time, for some fixed k
- So, it suffices to show how to construct a Boolean expression E(α) that says "there exists a string β such that the pair (α, β) is accepted by M in at most |α|^k steps".

Recall that the VERTEX-COVER problem takes as input a graph G and an integer k and asks if G has a vertex cover of size k, i.e. a subset $V_c \subseteq V$ such that every edge in E has at least one endpoint in V_c .

Note

- ► For *bipartite graphs* the vertex cover problem is in **P**
- In general VERTEX-COVER is in NP

The Vertex Cover Problem

In fact... Theorem: VERTEX-COVER is **NP**-complete.

Proof: We show that 3-SAT $\ensuremath{\boxtimes_{\mathrm{P}}}\ensuremath{\mathsf{VERTEX-COVER}}$



Building the NP-hardness reduction tree

- overview of reductions
- some selected examples
 - ▶ 3-SAT SP3-D-MATCHING
 - ▶ 3-D-MATCHING SPSUBSET-SUM













Some more examples

- ▶ 3-SAT SP3-D-MATCHING
- ▶ 3-D-MATCHING SPSUBSET-SUM