CS21 Decidability and Tractability

Lecture 22 February 28, 2014

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Outline

- · the class NP
 - NP-complete problems: subset sum
 - NP-complete problems: NAE-3-SAT, max-cut
- · the class co-NP
- the class NP ∩ coNP

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SUBSET-SUM is NP-complete

Theorem: the following language is NP-

complete: there is a subse

our reduction had SUBSET-SUM \neq {(S better produce superpolynomially large B (unless we want to

prove P=NP)

- Part 1: SUBSET-SU
 - Part 2: SUBSET-SUM is NP-hard.
 - · reduce from?

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• Proof:

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SUBSET-SUM is NP-complete

We are reducing from the language:

 $3SAT = \{ \phi : \phi \text{ is a 3-CNF formula that has a } \}$ satisfying assignment }

to the language:

SUBSET-SUM = $\{(S = \{a_1, a_2, a_3, ..., a_k\}, B) :$ there is a subset of S that sums to B}

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SUBSET-SUM is NP-complete

- $\varphi = (\mathbf{X}_1 \lor \mathbf{X}_2 \lor \neg \mathbf{X}_3) \land (\neg \mathbf{X}_1 \lor \mathbf{X}_4 \lor \mathbf{X}_3) \land \dots \land (\dots)$
- · Need integers to play the role of truth assignments
- For each variable x, include two integers in our set S:
 - $-x_i^{TRUE}$ and x_i^{FALSE}
- set B so that exactly one must be in sum

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SUBSET-SUM is NP-complete

 $x_1^{TRUE} = 1000...0$ $x_1^{FALSE} = 1000...0$ $x_2^{TRUE} = 0.100...0$ $x_2^{FALSE} = 0 1 0 0 ... 0$

· every choice of one from each $(x_i^{TRUE}, x_i^{FALSE})$ pair

sums to B

 $x_m^{TRUE} = 0 0 0 0 ... 1$ x_mFALSE = 0 0 0 0 ... 1 every subset that sums to B must choose one from each (x_i^{TRUE},x_i^{FALSE}) pair

= 1 1 1 1 ... 1

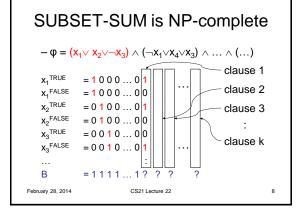
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SUBSET-SUM is NP-complete

- $\phi = (X_1 \lor X_2 \lor \neg X_3) \land (\neg X_1 \lor X_4 \lor X_3) \land \dots \land (\dots)$
- Need to force subset to "choose" at least one true literal from each clause
- · Idea:
 - add more digits
 - one digit for each clause
 - set B to force each clause to be satisfied.

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SUBSET-SUM is NP-complete

- B = 1 1 1 1 ... 1 ? ? ?
- if clause i is satisfied sum might be 1, 2, or 3 in corresponding column.
- want ? to "mean" ≥ 1
- solution: set ? = 3
- add two "filler" elements for each clause i:
- $-FILL1_i = 0 0 0 0 \dots 0 0 \dots 0 1 0 \dots 0$
- $FILL2_i = 0 0 0 0 \dots 0 0 \dots 0 1 0 \dots 0$

column for clause i

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SUBSET-SUM is NP-complete

- Reduction: m variables, k clauses
 - for each variable x_i:
 - x_i^{TRUE} has ones in positions k + i and {j : clause j includes literal x_i}
 - x_iFALSE has ones in positions k + i and {j : clause j includes literal ¬x_i}
 - for each clause i:
 - FILL1; and FILL2; have one in position i
 - bound B has 3 in positions 1...k and 1 in positions k+1...k+m

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SUBSET-SUM is NP-complete

- · Reduction computable in poly-time?
- YES maps to YES?
 - choose one from each (X_i^{TRUE},X_i^{FALSE}) pair corresponding to a satisfying assignment
 - choose 0, 1, or 2 of filler elements for each clause i depending on whether it has 3, 2, or 1 true literals
 - first m digits add to 1; last k digits add to 3

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SUBSET-SUM is NP-complete

- NO maps to NO?
 - at most 5 ones in each column, so no carries to worry about
 - first m digits of B force subset to choose exactly one from each (x_i^{TRUE}, x_i^{FALSE}) pair
 - last k digits of B require at least one true literal per clause, since can only sum to 2 using filler elements
 - resulting assignment must satisfy φ

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Not-All-Equal 3SAT

 $(x_1 \lor x_2 \lor \neg x_3) \land (\neg x_1 \lor x_4 \lor x_3) \land \dots \land (\dots)$

Theorem: the following language is NPcomplete:

NAE3SAT = $\{\phi: \phi \text{ is a 3-CNF formula for which there}$ exists a truth assignment in which every clause has at least 1 true literal and at least 1 false literal}

- Proof:
 - Part 1: NAE3SAT ∈ NP. Proof?
 - Part 2: NAE3SAT is NP-hard. Reduce from?

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NAE3SAT is NP-complete

 We are reducing from the language: CIRCUIT-SAT = {C : C is a Boolean circuit for which there exists a satisfying truth assignment)

to the language:

NAE3SAT = $\{\phi : \phi \text{ is a 3-CNF formula for which there}\}$ exists a truth assignment in which every clause has at least 1 true literal and at least 1 false literal}

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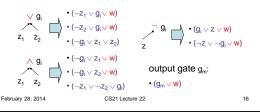
NAE3SAT is NP-complete

- Recall reduction to 3SAT
 - variables $x_1, x_2, ..., x_n$, gates $g_1, g_2, ..., g_m$

- produce clauses: not all true in a satisfying assignment $(\neg z_2 \lor g_i)$ $\begin{array}{c}
\bullet (g_i \lor z) \\
\bullet (\neg z \lor \neg g_i)
\end{array}$ • $(\neg g_i \lor z_1 \lor z_2)$ • $(\neg g_i \lor z_1)$ output gate g_m: • (g_m) $\bullet \; (\neg z_1 \vee \neg z_2 \vee g_i)$ February 28, 2014 CS21 Lecture 22 15

NAE3SAT is NP-complete

- modified reduction to NAE3SAT
 - variables x₁, x₂, ...,x_n, gates g₁, g₂, ..., g_m
 - produce clauses:



NAE3SAT is NP-complete

- · Does the reduction run in polynomial time?
- $(\neg z_2 \lor g_i \lor \mathbf{w})$ • $(\neg g_i \lor z_1 \lor z_2)$ • $(\neg g_i \lor z_1 \lor w)$

• $(\neg z_1 \lor g_i \lor w)$

- · YES maps to YES
- $(\neg g_i \lor z_2 \lor w)$
- already know how to get a satisfying assignment to the **BLUE** variables
- $(\neg z_1 \lor \neg z_2 \lor g_i)$ • $(g_i \lor z \lor w)$
- set w = FALSE
- $(\neg z \lor \neg g_i \lor w)$ • $(g_m \vee w)$

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NAE3SAT is NP-complete

NO maps to NO

assignment

- given NAE assignment A
- complement A' is a NAE
- -A or A' has w = FALSE
- must have TRUE BLUE
- satisfiable
- $(\neg g_i \lor z_1 \lor w)$ • $(\neg g_i \lor z_2 \lor w)$ • $(\neg z_1 \lor \neg z_2 \lor g_i)$

 $\bullet \ (\neg z_1 \vee g_i \vee w)$

• $(\neg z_2 \lor g_i \lor \mathbf{w})$

• ($\neg g_i \lor z_1 \lor z_2$)

- variable in every clause - we know this implies C
- $(g_i \lor z \lor w)$ • $(\neg z \lor \neg g_i \lor w)$

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• $(g_m \vee w)$

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MAX CUT

- Given graph G = (V, E)
 - a cut is a subset $S \subset V$
 - an edge (x, y) crosses the cut if $x \in S$ and $y \in V S$ or $x \in V S$ and $y \in S$
 - search problem:

find cut maximizing number of edges crossing the cut



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MAX CUT

- Given graph G = (V, E)
 - a cut is a subset $S \subset V$
 - an edge (x, y) crosses the cut if $x \in S$ and $y \in V S$ or $x \in V S$ and $y \in S$
 - search problem:

find cut maximizing number of edges crossing the cut



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MAX CUT

<u>Theorem</u>: the following language is NP-complete:

$$\label{eq:maxcut} \begin{split} \text{MAX CUT} = & \{(G = (V, E), k) : \text{there is a cut } S \subset V \text{ with at least } k \text{ edges crossing it} \} \end{split}$$

- · Proof:
 - Part 1: MAX CUT ∈ NP. Proof?
 - Part 2: MAX CUT is NP-hard.
 - reduce from?

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MAX CUT is NP-complete

We are reducing from the language:

 $NAE3SAT = \{\phi: \phi \text{ is a 3-CNF formula for which there}\\ \text{exists a truth assignment in which every clause has at}\\ \text{least 1 true literal and at least 1 false literal}\}$

to the language:

MAX CUT = $\{(G = (V, E), k) : \text{there is a cut } S \subset V \text{ with at least } k \text{ edges crossing it} \}$

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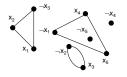
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MAX CUT is NP-complete

- · The reduction:
 - given instance of NAE3SAT (n nodes, m clauses):

 $(x_1 \!\!\vee x_2 \!\!\vee \!\!\neg x_3) \wedge (\neg x_1 \!\!\vee \!\! x_4 \vee \!\! x_5) \wedge \ldots \wedge (\neg x_2 \!\!\vee \!\! x_3 \vee \!\! x_3)$

- produce graph G = (V, E) with node for each literal

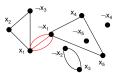


- triangle for each 3-clause
- parallel edges for each 2-clause

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MAX CUT is NP-complete



- triangle for each 3-clause
- parallel edges for each 2-clause
- if cut selects TRUE literals, each clause contributes 2 if NAE, and < 2 otherwise
- need to penalize cuts that correspond to inconsistent truth assignments
- add n_i parallel edges from x_i to ¬x_i (n_i = # occurrences)
 (repeat variable in 2-clause to make 3-clause for this calculation)

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