

CSE 6140/ CX 4140 Computational Science and Engineering ALGORITHMS

NP Completeness 3

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Based on slides by Prof. Ümit V. Çatalyürek

Reductions



- Reduction from A to B is showing that we can solve A using the algorithm that solves B
- We say that <u>problem A is easier than problem B</u>, (i.e., we write "A ≤ B")

Does it mean that the running time of A is less than B?

Not necessarily.

Polynomial Reductions



- Given two problems A, B, we say that A is polynomially reducible to B (A \leq_p B) if:
 - There exists a function f that converts the input of A to inputs of B in polynomial time
 - 2. $A(i) = YES \iff B(f(i)) = YES$

Summary



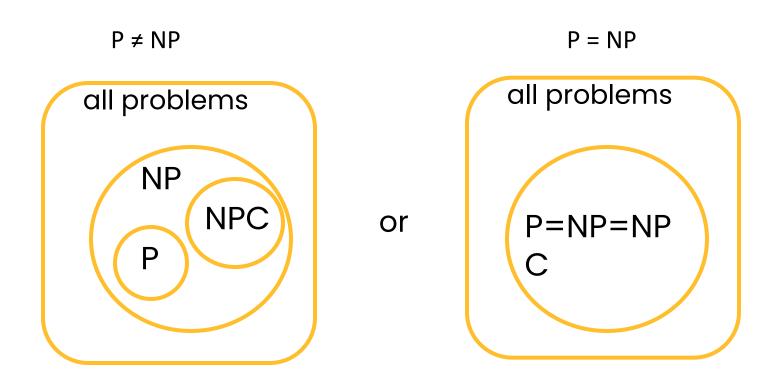
- . P
- Decision problems that can be solved in polynomial time
- Can be solved "efficiently"
- NP
 - Decision problems whose "YES" answer can be verified in polynomial time, if we already have the candidate solution

NP-complete

• The "hardest" problems in NP: a problem Y in NP with the property that for every problem X in NP, $X \leq_p Y$.

Possible Worlds

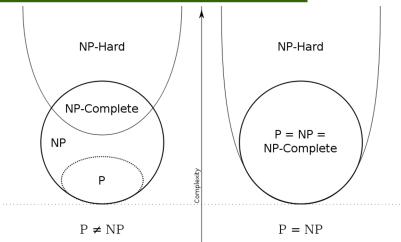




NPC: NP-complete

Revisit "Is P = NP?"





- Theorem. Suppose Y is an NP-complete problem. Y is solvable in poly-time if and only if P = NP.
- Pf. ← If P = NP then Y is in P. Hence Y can be solved in poly-time.
- Pf. ⇒ Suppose Y can be solved in poly-time.
 - Let X be any problem in NP. Then, we know that $X \le_p Y$ by definition of NP-complete and Y being NP-complete problem. Then we can solve X in poly-time by solving Y in poly-time. This implies any problem X in NP is also in P, i.e. NP \subseteq P.
 - We already know P ⊆ NP. Thus P = NP.

Implications of Polynomial-Time Reductions



- Purpose. Classify problems according to relative difficulty.
- Design algorithms. If $X \le_P Y$ and Y can be solved in polynomial-time, then X can also be solved in polynomial time.
- Establish intractability. If $X \le_P Y$ and X cannot be solved in polynomial-time, then Y cannot be solved in polynomial time.
- Establish equivalence. If $X \le_P Y$ and $Y \le_P X$, we use notation $X \equiv_P Y$.

up to cost of reduction

• Transitivity: if $X \leq_p Y$ and $Y \leq_p Z$, then $X \leq_p Z$

Reduction By Simple Equivalence



- Basic reduction strategies.
 - Reduction by simple equivalence.
 - Reduction from special case to general case.
- Reduction by encoding with gadgets.

Establishing NP-Completeness



- Recipe to establish NP-completeness of problem Y.
 - Step 1. Show that Y is in NP.
 - Describe how a potential solution/witness will be represented
 - Describe a procedure to check whether the potential witness is a correct solution to the problem instance, and argue that this procedure takes polynomial time
 - Step 2. Choose an NP-complete problem X.
 - Step 3. Prove that $X \leq_P Y$ (X is **poly-time reducible** to Y).
 - Describe a procedure f that converts the inputs i of X to inputs of Y in polynomial time
 - Show that the reduction is correct by showing that
 X(i) = YES ⇔ Y(f(i)) = YES (if and only if, proof in both directions)

P & NP-Complete Problems



Shortest simple path

- Given a graph G = (V, E) find a shortest path from a source to all other vertices
- Polynomial solution: O(VE)

Longest simple path

- Given a graph G = (V, E) find a longest path from a source to all other vertices
- NP-complete

P & NP-Complete Problems



Euler tour

- G = (V, E) a connected, directed graph find a cycle that traverses each
 edge of G exactly once (may visit a vertex multiple times)
- Polynomial solution O(E)

Hamiltonian cycle

- G = (V, E) a connected, directed graph find a cycle that visits each
 vertex of G exactly once
- NP-complete

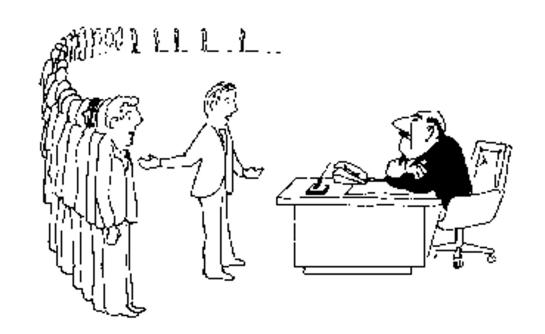
More Hard Computational Problems



- Aerospace engineering: optimal mesh partitioning for finite elements.
- Biology: protein folding.
- Chemical engineering: heat exchanger network synthesis.
- Civil engineering: equilibrium of urban traffic flow.
- Economics: computation of arbitrage in financial markets with friction.
- Electrical engineering: VLSI layout.
- Environmental engineering: optimal placement of contaminant sensors.
- Financial engineering: find minimum risk portfolio of given return.
- Game theory: find Nash equilibrium that maximizes social welfare.
- Genomics: phylogeny reconstruction.
- Mechanical engineering: structure of turbulence in sheared flows.
- Medicine: reconstructing 3-D shape from biplane angiocardiogram.
- Operations research: optimal resource allocation.
- Physics: partition function of 3-D Ising model in statistical mechanics.
- Politics: Shapley-Shubik voting power.
- Statistics: optimal experimental design.

Practical applications of NP-completeness





"I can't find an efficient algorithm, but neither can all these famous people."

[Garey & Johnson, Computers and Intractability: A Guide to the Theory of NP-Completeness, Freeman, 1979.]

Satisfiability Problem (SAT)



Satisfiability problem: given a logical expression Φ , find an assignment of values (F, T) to variables x_i that causes Φ to evaluate to T

$$\Phi = X_1 \vee \neg X_2 \wedge X_3 \vee \neg X_4$$

- boolean variables: take on values T or F
 - Ex: x, y
- literal: variable or negation of a variable
 - Ex: x, \neg x (also denoted by \bar{x})

Logical Operands



•
$$x = \{0,1\} \text{ or } \{F,T\}$$

Not

х	¬ x (negation)
0	1
1	0

And

x_{1}	X ₂	$x_1 \wedge x_2$ (AND)
1	1	1
1	0	0
0	1	0
0	0	0

Or

X ₁	X ₂	$x_1 \vee x_2$ (OR)
1	1	1
1	0	1
0	1	1
0	0	0

Satisfiability Problem (SAT)



• Satisfiability problem: given a logical expression Φ , find an assignment of values (F, T) to variables x_i that causes Φ to evaluate to T

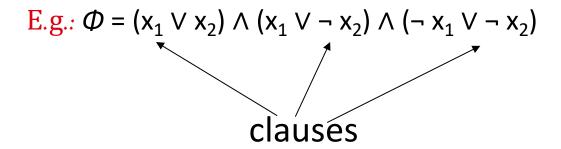
$$\Phi = \mathbf{x}_1 \vee \neg \mathbf{x}_2 \wedge \mathbf{x}_3 \vee \neg \mathbf{x}_4$$

- SAT is in NP: given a value assignment, check the Boolean logic of Φ evaluates to True (linear time)
- SAT was the first problem shown to be NP-complete! (Cook-Levin theorem)

CNF Satisfiability



- CNF is a special case of SAT
- Φ is in "Conjunctive Normal Form" (CNF)
 - "AND" of expressions (i.e., clauses)
 - Each clause contains only "OR"s of the variables and their negations



SAT-CNF is NP-Complete

In the following, SAT means SAT-CNF

Definition of 3SAT / 3CNF



- A subcase of CNF problem:
 - Contains three literals per clause
- E.g.:
 - $\Phi = (x1 \lor \neg x1 \lor \neg x2) \land (x3 \lor x2 \lor x4) \land (\neg x1 \lor \neg x3 \lor \neg x4)$
- Is 3SAT in NP?
 - Yes, because SAT is in NP. Also easy to prove it directly.
- Is 3SAT NP-complete?
 - Not obvious. It has a more regular structure, which can perhaps be exploited to get an efficient algorithm
 - In fact, 2SAT does have a polynomial time algorithm

Showing 3SAT is NP-Complete by Reduction



- (1) To show 3SAT is in NP:
 - A certificate is a truth (0/1) assignment to the variables
 - Certifier: check that each clause has at least one literal set to true according to the certificate
- (2) Choose SAT as known NP-complete problem
- (3) Describe a reduction from SAT inputs to 3SAT inputs
 - Computable in poly time
 - SAT input is satisfiable iff constructed 3SAT input is satisfiable
 - (3a) Transform I₁ (instance of SAT) into I₂ (instance of 3SAT) in polynomial time
 - (3b,3c) Prove that I₁ has a solution

 ⇔ I₂ has a solution

(3a) Reduction from SAT to 3SAT: $I_1 \rightarrow I_2$



- We are given an arbitrary CNF formula C = c₁Λ c₂ Λ ... Λ c_m over set of variables U, this is instance I₁
 - each c_i is a clause (disjunction of literals)
- We will replace each clause c_i with a set of clauses C_i', and may use some extra variables U_i' just for this clause
- Each clause in C_i' will have exactly 3 literals
- Transformed input will be conjunction of all the clauses in all the C_i', this is an instance I₂ of 3SAT
- New clauses are carefully chosen...





Let $c_i = z_1 \vee z_2 \vee ... \vee z_k$ (the z's are literals)

- Case 1: k = 1.
 - E.g. $c_i = z_1$
 - Use extra variables y_i¹ and y_i².
 - Replace clause c_i with 4 clauses:

$$\begin{array}{l} (z_1 \vee \overline{y_i^1} \vee y_i^2) \\ (z_1 \vee \overline{y_i^1} \vee \overline{y_i^2}) \end{array}$$
 z_1 can be replaced by the intersection of these 4 clauses
$$(z_1 \vee \overline{y_i^1} \vee \overline{y_i^2})$$

 Note that no matter what values we give the y variables, in one of the 4 clauses we will be forced to use z₁ to satisfy it





Let
$$c_i = z_1 \vee z_2 \vee ... \vee z_k$$

- Case 2: k = 2.
 - E.g. $c_i = z_1 V z_2$
 - Use extra variable y_i¹.
 - Replace c_i with 2 clauses:

$$(z_1 \lor z_2 \lor \overline{y_i^1})$$
$$(z_1 \lor z_2 \lor y_i^1)$$





Let
$$c_i = z_1 \vee z_2 \vee ... \vee z_k$$

- Case 3: k = 3.
 - No extra variables are needed.
 - Keep c_i:
 (z₁ V z₂ V z₃)





Let
$$c_i = z_1 \vee z_2 \vee ... \vee z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

(3a) Why is Reduction Poly Time?



- The running time of the reduction (the algorithm to compute the 3SAT formula C', given the SAT formula C) is proportional to the size of C'
- Rules for constructing C' are simple to calculate

(3a) Size of New Formula



- Original clause with 1 literal becomes 4 clauses with 3 literals each (1 to 12 literals conversion)
- Original clause with 2 literals becomes 2 clauses with 3 literals each (2 to 6 literals conversion)
- Original clause with 3 literals becomes 1 clause with 3 literals
- Original clause with k > 3 literals becomes k-2 clauses with 3 literals each (k to 3(k-2) literals conversion)
- So new formula C' is only a constant factor larger than the original formula
 - total L literals in formula C to cL literals in C', where c is a constant

(3bc) Correctness of Reduction



- Show that CNF formula C is satisfiable iff the 3-CNF formula C' constructed is satisfiable, i.e., sol(I₁) ⇔ sol(I₂)
- Step 3b (=>) Suppose original CNF formula C is satisfiable, i.e., I₁
 has a solution. That means it has a truth assignment A to the
 variables that make the formula C evaluate to true.
- Come up with a satisfying truth assignment for the reduced 3SAT formula C', i.e., a solution to instance I₂.
 - For variables in U, use same truth assignments as for C.
 - How to assign T/F to the new variables in C'?



(3b) Truth Assignment for New Var.: $sol(I_1) => sol(I_2)$

Let
$$c_i = z_1$$

- Case 1: k = 1.
- Use extra variables y_i¹ and y_i².
 - Replace c_i with 4 clauses:

$$(z_1 \vee \overline{y_i^1} \vee y_i^2)$$

 $(z_1 \vee y_i^1 \vee \overline{y_i^2})$
 $(z_1 \vee \overline{y_i^1} \vee \overline{y_i^2})$
 $(z_1 \vee y_i^1 \vee y_i^2)$

Since z_1 is true, it does not matter how we assign y_i^1 and y_i^2





Let
$$c_i = (z_1 \lor z_2)$$

- Case 2: k = 2.
 - Use extra variable y_i¹.
 - Replace c_i with 2 clauses:

$$(z_1 \lor z_2 \lor \overline{y_i^{1}})$$

 $(z_1 \lor z_2 \lor y_i^{1})$

Since either z_1 or z_2 is true, it does not matter how we assign y_i^1





Let
$$c_i = z_1 V z_2 V z_3$$

- Case 3: k = 3.
 - No extra variables are needed.
 - Keep c_i:
 (z₁ V z₂ V z₃)

No new variables.





Let
$$c_i = z_1 V z_2 V ... V z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

If first true literal is z₁ or z₂, set all y_i's to false: then all later clauses have a true literal





Let
$$c_i = z_1 V z_2 V ... V z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

If first true literal is z_{k-1} or z_k , set all y_i 's to true: then all earlier clauses have a true literal





Let
$$c_i = z_1 V z_2 V ... V z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

If first true literal is in between, set all earlier y_i's to true and set all later y_i's to false



(3c) Correctness of Reduction: $sol(I_2) => sol(I_1)$

- (<=) Suppose the newly constructed 3SAT formula C' is satisfiable, i.e., I₂ has a solution. We must show that the original SAT formula C is also satisfiable, i.e., I₁ has a solution.
- Use the same satisfying truth assignment for C as for C' (ignoring new variables).
- Show each original clause has at least one true literal in it.



Let
$$c_i = z_1$$

- Case 1: k = 1.
- Use extra variables y_i¹ and y_i².
 - Replace c_i with 4 clauses:

$$(z_1 \lor \overline{y_i^1} \lor y_i^2)$$

 $(z_1 \lor y_i^1 \lor \overline{y_i^2})$
 $(z_1 \lor \overline{y_i^1} \lor \overline{y_i^2})$
 $(z_1 \lor y_i^1 \lor y_i^2)$

For every assignment of y_i¹ and y_i², in order for all 4 clauses to have a true literal, z₁ must be true.



Let
$$c_i = (z_1 \vee z_2)$$

- Case 2: k = 2.
 - Use extra variable y_i¹.
 - Replace c_i with 2 clauses:

$$(z_1 \lor z_2 \lor \overline{y_i^{1}})$$

$$(z_1 \lor z_2 \lor y_i^{1})$$

For either assignment of y_i^1 , in order for both clauses to have a true literal, z_1 or z_2 must be true.



Let
$$c_i = z_1 \vee z_2 \vee z_3$$

- Case 3: k = 3.
 - No extra variables are needed.
 - Keep c_i:
 (z₁ V z₂ V z₃)

No new variables.



Let
$$c_i = z_1 V z_2 V ... V z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

$$(\mathbf{z_1} \vee \mathbf{z_2} \vee \mathbf{y_i^1})$$

$$(\mathbf{\overline{y_i^1}} \vee \mathbf{z_3} \vee \mathbf{y_i^2})$$

$$(\mathbf{\overline{y_i^2}} \vee \mathbf{z_4} \vee \mathbf{y_i^3})$$
...

Suppose in contradiction all z_i's are false.

. . .

$$(\overline{y_i^{k-5}} \vee \mathbf{z_{k-3}} \vee y_i^{k-4})$$

$$(\overline{y_i^{k-4}} \vee \mathbf{z_{k-2}} \vee y_i^{k-3})$$

$$(\overline{y_i^{k-3}} \vee \mathbf{z_{k-1}} \vee \mathbf{z_k})$$



Let
$$c_i = z_1 \vee z_2 \vee ... \vee z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

$$(\mathbf{z}_1 \vee \mathbf{z}_2 \vee \mathbf{y}_i^1)$$

$$(\mathbf{\overline{y}_i^1} \vee \mathbf{z}_3 \vee \mathbf{y}_i^2)$$

$$(\mathbf{\overline{y}_i^2} \vee \mathbf{z}_4 \vee \mathbf{y}_i^3)$$
...

Suppose in contradiction all z_i 's are false.

Then y_i^1 must be true.

 $(\overline{y_i^{k-5}} \vee \underline{z_{k-3}} \vee y_i^{k-4})$

$$(\overline{y_i^{k-4}} \vee \underline{z_{k-2}} \vee y_i^{k-3})$$

$$(\overline{y_i^{k-3}} \lor z_{k-1} \lor z_k)$$



Let
$$c_i = z_1 \vee z_2 \vee ... \vee z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

$$(z_1 \lor z_2 \lor y_i^1)$$

$$(\overline{y_i^1} \lor z_3 \lor y_i^2)$$

$$(\overline{y_i^2} \lor z_4 \lor y_i^3)$$
...

Suppose in contradiction all z_i's are false.

Then y_i¹ must be true.

Then y_i² must be true...

. . .

$$(\overline{y_i^{k-5}} \vee z_{k-3} \vee y_i^{k-4})$$

$$(\overline{y_i^{k-4}} \vee z_{k-2} \vee y_i^{k-3})$$

$$(\overline{y_i^{k-3}} \vee z_{k-1} \vee z_k)$$



Let
$$c_i = z_1 \vee z_2 \vee ... \vee z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

$$(z_1 \lor z_2 \lor y_i^1)$$

 $(\overline{y_i^1} \lor z_3 \lor y_i^2)$
 $(\overline{y_i^2} \lor z_4 \lor y_i^3)$

Suppose in contradiction all z_i 's are false.

Then y_i^1 must be true.

Then y_i^2 must be true...

$$(\overline{y_i^{k-5}} \vee z_{k-3} \vee y_i^{k-4})$$

$$(\overline{y_i^{k-4}} \vee z_{k-2} \vee y_i^{k-3})$$

$$(\overline{y_i^{k-3}} \vee z_{k-1} \vee z_k)$$



Let
$$c_i = z_1 \vee z_2 \vee ... \vee z_k$$

- Case 4: k > 3.
 - Use extra variables y_i¹, ..., y_i^{k-3}.
 - Replace c_i with k-2 clauses:

$$(z_1 \lor z_2 \lor y_i^1)$$

 $(\overline{y_i^1} \lor z_3 \lor y_i^2)$
 $(\overline{y_i^2} \lor z_4 \lor y_i^3)$

Suppose in contradiction all z_i's are false. Then y_i¹ must be true. Then y_i² must be true...

So the last clause is False

. . .

$$(\overline{y_i^{k-5}} \lor z_{k-3} \lor y_i^{k-4})$$

$$(\overline{y_i^{k-4}} \lor z_{k-2} \lor y_i^{k-3})$$

$$(\overline{y_i^{k-3}} \lor z_{k-1} \lor z_k)$$

Conclusion

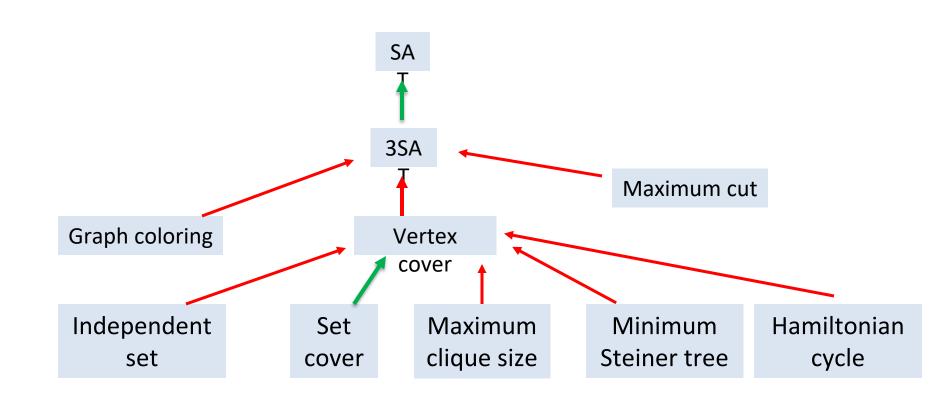


- (1) 3SAT is in NP
- (2) We know that SAT is NPC, we want to prove that 3SAT is more difficult than SAT, hence SAT ≤_P 3SAT
- (3a) Take an instance I₁ of SAT, transform it in polynomial time into an instance I₂ of 3SAT
- (3b) Show that if I₁ has a solution, then I₂ has a solution
- (3c) Show that if I_2 has a solution, then I_1 has a solution
- 3SAT is NP-complete! This is your very first NP-completeness proof. Now you can do reductions from 3SAT.
- (All pbs in NP) $\leq_p SAT \leq_p 3SAT$

Examples of NP-complete problems

Summary of some NPC problems





find more NP-complete problems in

- http://en.wikipedia.org/wiki/List of NP-complete problems
- Garey-Johnson: Computers and Intractability: A Guide to the Theory of NP-Completeness

Genres of NP-complete problems



- Six basic genres of NP-complete problems and paradigmatic examples.
 - Packing problems: SET-PACKING, INDEPENDENT SET.
 - Covering problems: SET-COVER, VERTEX-COVER.
 - Constraint satisfaction problems: SAT, 3-SAT.
 - Sequencing problems: HAMILTONIAN-CYCLE, TSP.
 - Partitioning problems: 3-COLOR, 3D-MATCHING.
 - Numerical problems: 2-PARTITION, SUBSET-SUM, KNAPSACK.