

## CS 440 More on Reduction, NP and NP-Complete

- Polynomial Reductions

- Definition: Polynomial Turing Reduction

- Let  $A$  and  $B$  be two problems. We say that  $A$  is *polynomially Turing reducible* to  $B$ , denoted  $A \leq_T^P B$ , if there exists an algorithm for solving  $A$  in a time that would be polynomial if we could solve arbitrary instances of problem  $B$  at unit cost.
- In other words, the algorithm for solving problem  $A$  may make whatever use it chooses of an imaginary algorithm that can solve problem  $B$  at unit cost.

- Definition: Polynomial Turing Equivalence

- $A$  and  $B$  are *polynomially Turing equivalent*, denoted  $A \equiv_T^P B$ , if  $A \leq_T^P B$  and  $B \leq_T^P A$ .

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- Example: Hamiltonian cycle problem (HAM) and the Hamiltonian cycle decision problem (HAMD).

- It turns out that it is not significantly harder to find a Hamiltonian cycle than to decide if a graph is Hamiltonian.

- Theorem:  $HAM \equiv_T^P HAMD$

- Proof: (must show two things: 1.  $HAMD \leq_T^P HAM$  and 2.  $HAM \leq_T^P HAMD$ )

1.  $HAMD \leq_T^P HAM$

```
function HamD(G: graph)
```

```
   $\sigma \leftarrow Ham(G)$ 
```

```
  if  $\sigma$  defines a Hamiltonian cycle in  $G$  then
```

```
    return true
```

```
  else
```

```
    return false
```

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## 2. $HAM \leq_T^P HAMD$

```
function Ham( $\langle N, A \rangle$ : graph)
  if HamD( $\langle N, A \rangle$ ) = false then
    return "no solution"
  for each  $e \in A$  do
    if HamD( $\langle N, A - \{e\} \rangle$ ) then
       $A \leftarrow A - \{e\}$ 
   $\sigma \leftarrow$  sequence of nodes obtained by following the unique cycle remaining
  return  $\sigma$ 
```

From 1 and 2,  $HAM \equiv_T^P HAMD$ .

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- Theorem: Consider two problems  $A$  and  $B$ . If  $A \leq_T^P B$  and if  $B$  can be solved in polynomial time then  $A$  can also be solved in polynomial time.
  - It follows that a polynomial time algorithm exists to find a Hamiltonian cycle if and only if a polynomial time algorithm exists to decide if a graph is Hamiltonian.
  - This is typical of many interesting problems which are polynomially equivalent to a similar decision problem
    - *Decision reducible*: If a problem interests you is not a decision problem, you probably can find a similar decision problem that is polynomially equivalent.

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- Polynomial Many-to-One Reductions

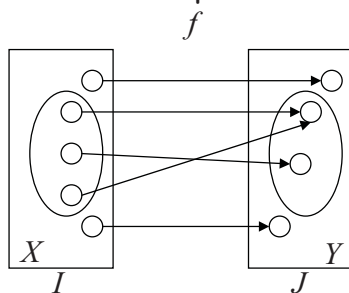
- Definition: Polynomial many-to-one reduction

- Let  $X$  and  $Y$  be two decision problems defined on sets of instances  $I$  and  $J$ , respectively. Problem  $X$  is *polynomially many-to-one reducible* to problem  $Y$ , denoted  $X \leq_m^P Y$ , if there exists a reduction function  $f: I \rightarrow J$  computable in polynomial time such that  $x \in X$  if and only if  $f(x) \in Y$  for any instance  $x \in I$  of problem  $X$ .

- Definition: Polynomial many-to-one equivalence

- When  $X \leq_m^P Y$  and  $Y \leq_m^P X$ ,  $X$  and  $Y$  are *polynomially many-one equivalent*, denoted  $X \equiv_m^P Y$

- In other words, the reduction function maps all yes-instances of problem  $X$  onto yes-instances of problem  $Y$  and all no-instances of problem  $X$  onto no-instances of problem  $Y$ .



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- Theorem: If  $X$  and  $Y$  are two decision problems such that  $X \leq_m^P Y$  then  $X \leq_T^P Y$ .

- Proof: Imagine solutions to  $Y$  can be obtained at unit cost by a call on  $\text{DecideY}$  and let  $f$  be the reduction function between  $X$  and  $Y$  computable in polynomial time. Consider the following algorithm:

```
function DecideX(x)
  y ← f(x)
  if DecideY(y) then
    return true
  else
    return false
```

By definition of the reduction function, this algorithm solves problem  $X$ . Because the reduction function is computable in polynomial time, it solves problem  $X$  in polynomial time since calls to  $\text{DecideY}$  can be counted at unit cost.

- Sometimes if we want to prove that  $X \leq_T^P Y$ , proving  $X \leq_m^P Y$  is sufficient.

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- Theorem:  $HAMD \leq_T^P$  TSPD (Traveling Salesperson Decision problem)
  - Proof:
    - Let  $G = \langle N, A \rangle$  be a graph with  $n$  nodes. We would like to decide if it is Hamiltonian.
    - Define  $f(G)$  as the instance of TSPD consisting of the complete graph  $H = \langle N, N \times N \rangle$ ,
    - Define the cost function for an edge in  $H$ :  $c(u, v) = \begin{cases} 1, & \text{if } (u, v) \in A \\ 2, & \text{otherwise} \end{cases}$   
and the bound  $L = n$ .
    - Any Hamiltonian cycle in  $G$  translates into a tour in  $H$  that has cost exactly  $n$ . On the other hand, if there is no Hamiltonian cycle in  $G$ , any tour in  $H$  must use at least one edge of cost 2, and thus be of total cost at least  $n+1$ .
    - Therefore,  $G$  is a yes-instance of HAMD if and only if  $f(G) = \langle H, c, L \rangle$  is a yes-instance of TSPD.
    - This proves that  $HAMD \leq_m^P$  TSPD  $\rightarrow$   $HAMD \leq_T^P$  TSPD

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- NP-Complete Problems
  - Definition: A decision problem  $X$  is NP-complete if:
    - $X \in NP$ ; and
    - $Y \leq_T^P X$  for every problem  $Y \in NP$  (i.e.,  $X$  is NP-Hard)
- Theorem: Let  $X$  be an NP-complete problem. Consider a decision problem  $Z \in NP$  such that  $X \leq_T^P Z$ . Then  $Z$  is also NP-complete.
  - Proof:  $Z$  must meet the two conditions of the definition of NP-completeness.
    - $Z \in NP$ . It is in the statement of the theorem.
    - $Z$  is NP-Hard (i.e.,  $Y \leq_T^P Z$  for every problem  $Y \in NP$ )  
Consider an arbitrary  $Y \in NP$ . Since  $X$  is NP-complete and  $Y \in NP$ , it follows that  $Y \leq_T^P X$ .  
We know by the statement of the theorem that  $X \leq_T^P Z$ .  
By *transitivity* of polynomial reductions it follows  $Y \leq_T^P Z$ .

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- To prove that Z is NP-complete,
  - We prove that  $Z \in NP$  by verifying a given solution to Z in polynomial time if it is not in the statement of the theorem.
  - We choose an *appropriate* problem from the pool of several thousand problems already known to be NP-complete and show that it is polynomially reducible to Z.
- The previous statement works nicely once the process is underway (i.e., once at least one NP-complete problem has been identified)
  - Q: How do we find the *FIRST* NP-complete problem? (What is it?)
  - Good news: We don't have to find it. But we have to know what it is.
- Satisfiability problem (SAT)
  - Definition: A Boolean formula is *satisfiable* if there exists at least one way of assigning values to its variables so as to make it true.
  - Example:  $(\neg(p \vee q)) \vee (p \wedge q)$  YES      $\neg p \wedge (p \vee q) \wedge \neg q$  NO
  - $SAT \in NP$

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- SAT-CNF
  - A literal is either a Boolean variable or its negation.
  - A clause is a literal or a disjunction of literals.
  - A Boolean formula is in *conjunctive normal form* (CNF) if it is a clause or a conjunction of clauses.
  - It is in k-CNF for some positive integer k if it is composed of clauses, each of which contains at most k literals.
    - Examples:
 

$(p \vee \neg q \vee r) \wedge (\neg p \vee q \vee r) \wedge q \wedge \neg r$	CNF?	k = ?
$(p \vee q \wedge r) \wedge (\neg p \vee q \wedge (q \vee r))$	CNF?	k = ?
  - SAT-CNF is a restriction of SAT to Boolean formulas in CNF.
  - For any positive k, SAT-k-CNF is the restriction of SAT-CNF to Boolean formulas in k-CNF.
- Theorem (due to Stephen Cook, 1971):
  - SAT-CNF is NP-complete.

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- NP-Completeness Proofs

- Theorem: SAT is NP-complete

- Proof:

- 1.  $SAT \in NP$ .

Given an assignment of variables, we can verify the Boolean formula is true or false in polynomial time.

- 2. (must show  $SAT-CNF \leq_T^P SAT$ )

Boolean formulas in CNF are special cases of general Boolean formulas

→  $SAT-CNF \leq_T^P SAT$

- Theorem: SAT-3-CNF is NP-complete

- Proof:

- 1.  $SAT-3-CNF \in NP$ .

Given an assignment of variables, we can verify the Boolean formula is true or false in polynomial time.

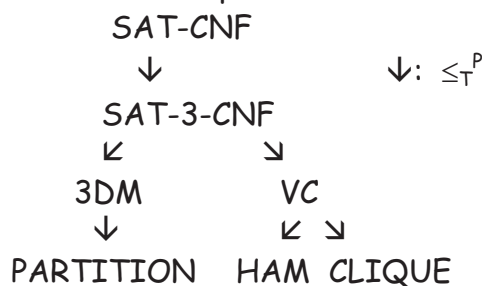
- 2. (must show  $SAT-CNF \leq_T^P SAT-3-CNF$ )

Proof will be given in the class if time allows.

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- More basic known NP-complete problems for NP-completeness proofs

- Basic known NP-complete problems that can be used to show some other problem is NP-complete



- VC (Vertex Cover): Given a graph and an integer K, is there a set of less than K vertices that touches all the edges?

Instance: A graph  $G = (V, E)$  and a positive integer  $K \leq |V|$

Question: Is there a vertex cover of size K or less for G?

(i.e., Is there a subset  $V' \subseteq V$  such that  $|V'| \leq K$  and, for every edge  $\{u, v\} \in E$ , at least one of u and v belongs to  $V'$ ?)

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- CLIQUE  
Instance: A graph  $G = (V, E)$  and a positive integer  $J \leq |V|$   
Question: Does  $G$  contain a clique of size of  $J$  or more?  
(i.e., Is there a complete subgraph of  $G$  that has at least  $J$  vertices?)
  
- 3DM: 3-Dimensional Matching  
Instance: A set  $M \subseteq W \times X \times Y$ , where  $W, X,$  and  $Y$  are disjoint sets and  $|W| = |X| = |Y| = q$   
Question: Does  $M$  contain a matching?  
(i.e., Is there a subset  $M' \subseteq M$  such that  $|M'| = q$  and no two elements of  $M'$  agree in any coordinate.)  
Note that 2DM is not NP-complete.
  
- PARTITION: Given a set of integers, can they be divided into two sets whose sum is equal?  
Instance: A finite set  $A$  and a *size*  $s(a) \in \mathbb{Z}^+$ , for all  $a \in A$   
Question: Is there a subset  $A' \subseteq A$  such that  $\sum_{a \in A'} s(a) = \sum_{a \in A-A'} s(a)$ ?

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- Example

- Knapsack problem  
Instance: A finite set  $U$ ,  
a weight  $w(u) \in \mathbb{Z}^+$  and a value  $v(u) \in \mathbb{Z}^+$ , for all  $u \in U$ ,  
a capacity  $W \in \mathbb{Z}^+$ , and a value goal  $K \in \mathbb{Z}^+$ .  
Question: Is there a subset  $U' \subseteq U$  such that  
 $\sum_{u \in U'} w(u) \leq W$  and  $\sum_{u \in U'} v(u) \geq K$ ?
  
- Theorem: Knapsack problem is NP-complete
  - Proof:
    1. Show Knapsack problem  $\in$  NP.
    2. Reduce a well-known NP-complete problem to Knapsack problem. (But, which one?)

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- NP-Hard Problems

- A problem  $X$  is *NP-hard* if there is an NP-complete problem  $Y$  that can be polynomially Turing reduced to it (i.e.,  $Y \leq_T^P X$ )
- Note that any polynomial-time algorithm for  $X$  would translate into one for  $Y$ . Since  $Y$  is NP-complete, this would imply that  $P = NP$ .
  - Under the assumption of  $P \neq NP$ , no NP-hard problem can be solved in polynomial time.
- Note that NP-hard contains non-decision problems.
  - Example:
    - TSP
    - 0-1 knapsack problem, time  $\in \Theta(nW)$ , is it polynomial? What if  $W \in \Theta(2^n)$ ?
- Further note that there are decision problems that are NP-hard but are believed to not be in NP and thus not in NP-complete.
  - Example:
    - COLE: exact coloring, given a graph  $G$  and an integer  $k$ , can  $G$  be painted with  $k$  colors but no less?