

# CHAPTER 8 BEYOND POLYNOMIAL RUNNING TIMES

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### **Outline**

- Content:
  - Polynomial-time reduction
  - NP-completeness
- Reading:
  - Chapter 8

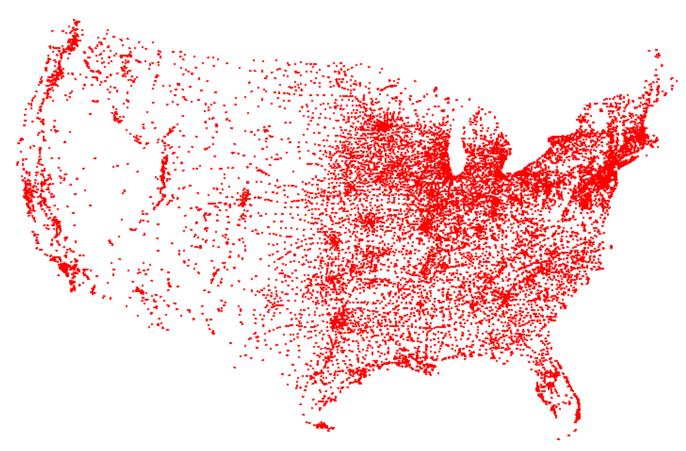
## Easy vs. Hard

- Q: Which problems will we be able to solve in practice?
- A working definition.
  - [Cobham 1964, Edmonds 1965, Rabin 1966] Those with polynomial-time algorithms.
- Desiderata: Classify problems according to those that can be solved in polynomial-time and those that cannot.
- Provably requires exponential-time.
- Frustrating news: Huge number of fundamental problems have defined classification for decades.
- Chapter 8: Show that these fundamental problems are computationally equivalent and appear to be different manifestations of one really hard problem.

## **Decision & Optimization Problems**

- Decision problems: those having yes/no answers.
  - MST: Given a graph G=(V, E) and a bound K, is there a spanning tree with a cost at most K?
  - TSP: Given a set of cities, distance between each pair of cities, and a bound B, is there a route that starts and ends at a given city, visits every city exactly once, and has total distance at most B?
- Optimization problems: those finding a legal configuration such that its cost is minimum (or maximum).
  - MST: Given a graph G=(V, E), find the cost of a minimum spanning tree of G.
  - TSP: Given a set of cities and that distance between each pair of cities, find the distance of a "minimum route" starts and ends at a given city and visits every city exactly once.
- Could apply binary search on a decision problem to obtain solutions to its optimization problem.
- Class NP is associated with decision problems.

## **Traveling Salesman Problem (TSP) (1/2)**



All 13,509 cities in US with a population of at least 500

## Traveling Salesman Problem (TSP) (2/2)



Optimal TSP tour

## **Complexity Classes**

- Developed by S. Cook and R. Karp in early 1970.
- The class P: class of problems that can be solved in polynomial time in the size of input.
- The class NP (Nondeterministic Polynomial): class of problems that can be verified in polynomial time in the size of input.
  - P=NP?
- The class NP-complete (NPC): A problem Y in NP with the property that for every problem X in NP,  $X \leq_p Y$ .
- Theorem: Suppose Y is NPC, then Y is solvable in polynomial time iff P = NP.
  - Any NPC problem can be solved in polynomial time ⇒ All problems in NP can be solved in polynomial time.
- Fundamental question: Do there exist "natural" NPC problems?

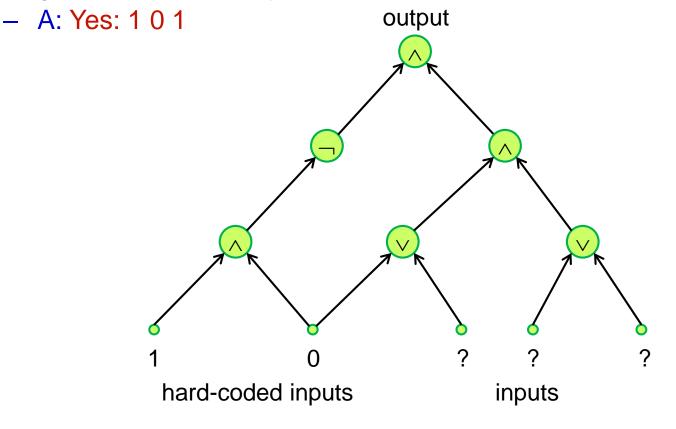
## Verification Algorithm and Class NP

- Verification algorithm: a 2-argument algorithm A, where one argument is an input string x and the other is a binary string y (called a certificate). A verifies x if there exists y s.t. A answers "yes."
- Ex: The Traveling Salesman Problem (TSP)
  - Instance: a set of *n* cities, distance between each pair of cities, and a bound *B*.
  - Question: is there a route that starts and ends at a given city, visits every city exactly once, and has total distance ≤ B?
- Is TSP ∈ NP?
- Need to check a solution in polynomial time.
  - Guess a tour (certificate).
  - Check if the tour visits every city exactly once.
  - Check if the tour returns to the start.
  - Check if total distance  $\leq B$ .
- All can be done in O(n) time, so TSP  $\in$  NP.

## The First Proved NPC: Circuit Satisfiability

#### CIRCUIT-SAT:

 Q: Given a combinational circuit built out of AND, OR, and NOT gates, is there a way to set the circuit inputs so that the output is 1?



## **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.
- Pop culture: Minesweeper consistency.
- Statistics: optimal experimental design.

## Polynomial-Time Reduction (1/2)

- Desiderata: Suppose we could solve Y in polynomialtime. What else could we solve in polynomial time?
- Reduction: Problem X polynomial reduces to problem Y if given an arbitrary instance x of problem X, we can construct an input y to problem Y in polynomial time such that x is a yes instance to X iff y is a yes instance of Y.
  - Notation: X ≤<sub>P</sub> Y.

#### • Remarks:

- The algorithm for Y is viewed as a black box.
- We pay for polynomial time to write down instances sent to this black box

## Polynomial-Time Reduction (2/2)

- Purpose: Classify problems according to relative difficulty.
- Design algorithms: If X ≤<sub>P</sub> Y and Y can be solved in polynomial-time, then X can also be solved in polynomial time.
  - Bipartite matching ≤<sub>P</sub> Network flow
- 2. Establish intractability: If  $X \leq_P Y$  and X cannot be solved in polynomial-time, then Y cannot be solved in polynomial time.
  - Hamiltonian cycle ≤<sub>P</sub> Travelling salesman
- 3. Establish equivalence: If  $X \leq_P Y$  and  $Y \leq_P X$ ,  $X \equiv_P Y$ .
  - Up to cost of reduction

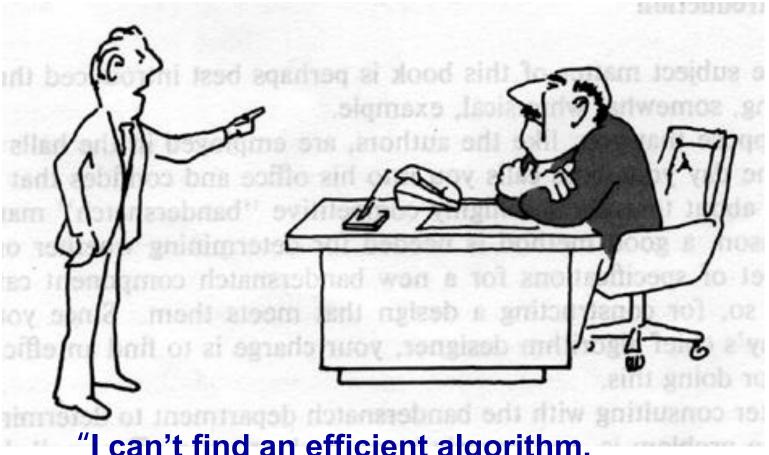
## Coping with a "Tough" Problem: Trilogy I



"I can't find an efficient algorithm.

I guess I'm just too dumb."

## Coping with a "Tough" Problem: Trilogy II



"I can't find an efficient algorithm, because no such algorithm is possible!"

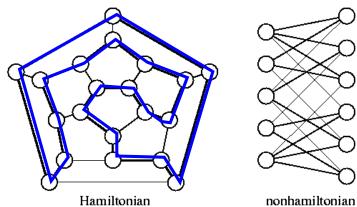
## Coping with a "Tough" Problem: Trilogy III



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

## Polynomial Reduction: HC ≤<sub>P</sub> TSP

- The Hamiltonian Circuit Problem (HC)
  - **Instance:** an undirected graph G = (V, E).
  - Question: is there a cycle in G that includes every vertex exactly once?
- TSP: The Traveling Salesman Problem
- Claim:  $HC \leq_P TSP$ .
  - 1. Define a function *f* mapping **any** HC instance into a TSP instance, and show that *f* can be computed in polynomial time.
  - 2. Prove that G has an HC iff the reduced instance has a TSP tour with distance  $\leq B$  ( $x \in HC \Leftrightarrow f(x) \in TSP$ ).

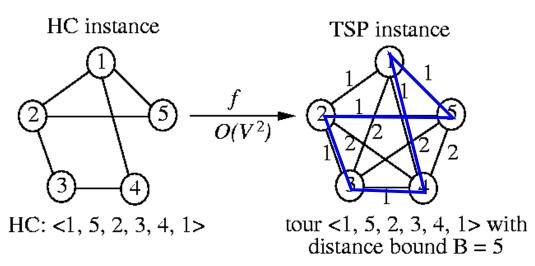


## $HC \leq_P TSP$ : Step 1

- Define a reduction function f for  $HC \leq_P TSP$ .
  - Given an HC instance G = (V, E) with n vertices
    - Create a set of n cities labeled with names in V.
    - Assign distance between u and v

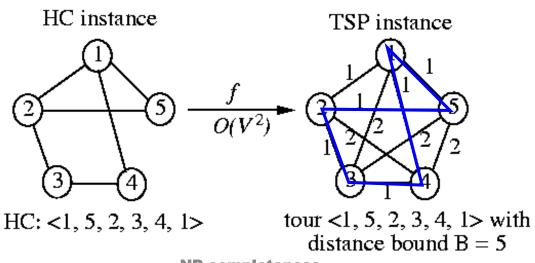
$$d(u,v) = \begin{cases} 1, & \text{if } (u,v) \in E, \\ 2, & \text{if } (u,v) \notin E. \end{cases}$$

- Set bound B = n.
- f can be computed in  $O(V^2)$  time.



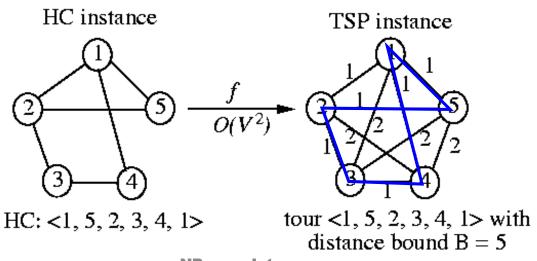
## HC ≤<sub>P</sub> TSP: Step 2

- G has a HC iff the reduced instance has a TSP with distance ≤ B.
  - $-x \in HC \Rightarrow f(x) \in TSP.$ 
    - Suppose the HC is  $h = \langle v_1, v_2, ..., v_n, v_1 \rangle$ . Then, h is also a tour in the transformed TSP instance.
    - The distance of the tour h is n = B since there are n consecutive edges in E, and so has distance 1 in f(x).
    - Thus,  $f(x) \in TSP(f(x))$  has a TSP tour with distance  $\leq B$ ).



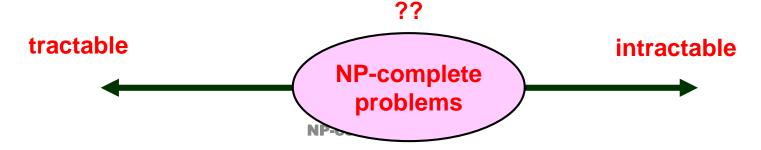
## $HC \leq_P TSP$ : Step 2 (cont'd)

- G has a HC iff the reduced instance has a TSP with distance ≤ B.
  - $f(x) \in \mathsf{TSP} \Rightarrow x \in \mathsf{HC}.$ 
    - Suppose there is a TSP tour with distance  $\leq n = B$ . Let it be  $\langle v_1, v_2, ..., v_n, v_1 \rangle$ .
    - Since distance of the tour  $\leq n$  and there are n edges in the TSP tour, the tour contains only edges in E.
    - Thus,  $\langle v_1, v_2, ..., v_n, v_1 \rangle$  is a Hamiltonian cycle ( $x \in HC$ ).



## **NP-Completeness**

- Definition: A decision problem L (a language L ⊆ {0, 1}\*)
  is NP-complete (NPC) if
  - 1.  $L \in NP$ , and
  - 2.  $L' \leq_{\mathbf{P}} L$  for every  $L' \in \mathbb{NP}$ .
- NP-hard: If L satisfies property 2, but not necessarily property 1, we say that L is NP-hard.
- Suppose  $L \in NPC$ . P=NP?
  - If  $L \in P$ , then there exists a polynomial-time algorithm for every  $L' \in NP$  (i.e., P = NP).
  - If  $L \notin P$ , then there exists no polynomial-time algorithm for any  $L' \in NPC$  (i.e.,  $P \neq NP$ ).



## **Proving NP-Completeness**

- Theorem: A decision problem L (a language L ⊆ {0, 1}\*)
  is NP-complete (NPC) if
  - 1.  $L \in NP$ , and
  - 2.  $L' \leq_{\mathbf{P}} L$  for an  $L' \in \mathsf{NPC}$ .
- Five steps for proving that L is NP-complete:
  - 1. Prove  $L \in NP$ .
  - 2. Select a known NP-complete problem L'.
  - 3. Construct a reduction *f* transforming **every** instance of *L*' to an instance of *L*.
  - 4. Prove that *f* is a polynomial-time transformation.
  - 5. Prove that  $x \in L'$  iff  $f(x) \in L$  for all  $x \in \{0, 1\}^*$ .

