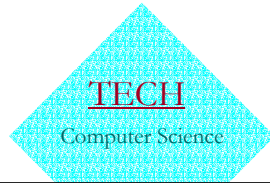


## NP-Complete Problems

- Problems
  - **Abstract Problems**
    - > Decision Problem, Optimal value, Optimal solution
  - **Encodings**
    - > //Data Structure
  - **Concrete Problem**
    - > //Language
- Class of Problems
  - P
  - NP
  - **NP-Complete**
    - > NP-Completeness Proofs
- Solving hard problems
  - **Approximation Algorithms**



## Abstract Problems

- a formal notion of what a “problem” is
- high-level description of a problem
- We define an *abstract problem*  $Q$  to be
  - a binary relation on
  - a set  $I$  of problem *instances*, and
  - a set  $S$  of problem *solutions*.
  - $Q \in I \times S$
- Three Kinds of Problems
  - **Decision Problem**
    - > e.g. Is there a solution better than some given bound?
  - **Optimal Value**
    - > e.g. What is the value of a best possible solution?
  - **Optimal Solution**
    - > e.g. Find a solution that achieves the optimal value.

## Encodings

- // Data Structure
- describing abstract problems (for solving by computers)
- in terms of data structure or binary strings
- An *encoding* of a set  $S$  of abstract objects is
  - a mapping  $e$  from  $S$  to the set of binary strings.
- Encoding for Decision problems
  - Problem instances,  $e : I \rightarrow \{0, 1\}^*$
  - Solution,  $e : S \rightarrow \{0, 1\}$
- “Standard” encoding
  - computing time may be a function of encoding
    - > // The size of the input (the number of bit to represent one input)
  - polynomially related encodings
  - assume encoding in a reasonable concise fashion

## Concrete Problem

- problem instances and solutions are represented in data structure or binary strings
- // Language (in formal-language framework)
- We call a problem whose instance set (and solution set) is the set of binary strings a *concrete problem*.
- Computer algorithm solves concrete problems!
  - solves a concrete problem in time  $O(T(n))$
  - if provided a problem instance  $i$  of length  $n = |i|$ ,
  - the algorithm can produce the solution
  - in a most  $O(T(n))$  time.
- A concrete problem is *polynomial-time solvable*
  - if there exists an algorithm to solve it in time  $O(n^k)$
  - for some constant  $k$ . (also called polynomially bounded)

## Class of Problems

- // What makes a problem hard?
- // Make simple: classify decision problems
- Definition: The class  $P$ 
  - $P$  is the class of decision problems that are polynomially bounded.
  - > // there exist a deterministic algorithm
- Definition: The class  $NP$ 
  - $NP$  is the class of decision problems for which there is a polynomially bounded non-deterministic algorithm.
  - > The name  $NP$  comes from “Non-deterministic Polynomially bounded.”
  - > // there exist a non-deterministic algorithm
- Theorem:  $P \subseteq NP$

## The Class NP

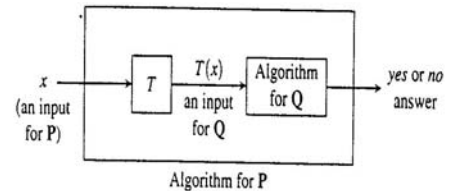
- NP is a class of decision problems for which
  - a given proposed solution (called certificate) for
  - a given input
  - can be checked quickly (in polynomial time)
  - to see if it really is a solution.
- A non-deterministic algorithm
  - The non-deterministic “guessing” phase.
    - > Some completely arbitrary string  $s$ , “proposed solution”
    - > each time the algorithm is run the string may differ
  - The deterministic “verifying” phase.
    - > a deterministic algorithm takes the input of the problem and the proposed solution  $s$ , and
    - > return value true or false
  - The output step.
    - > If the verifying phase returned true, the algorithm outputs yes. Otherwise, there is no output.

## The Class NP-Complete

- A problem Q is *NP-complete*
  - if it is in NP and
  - it is NP-hard.
- A problem Q is *NP-hard*
  - if every problem in NP
  - is reducible to Q.
- A problem P is *reducible* to a problem Q if
  - there exists a polynomial reduction function T such that
    - > For every string x,
    - > if x is a yes input for P, then T(x) is a yes input for Q
    - > if x is a no input for P, then T(x) is a no input for Q.
    - > T can be computed in polynomially bounded time.

## Polynomial Reductions

- Problem P is reducible to Q
  - $P \leq_p Q$
  - Transforming inputs of P
    - > to inputs of Q
- Reducibility relation is transitive.



## Circuit-satisfiability problem is NP-Complete

- Circuit-satisfiability problem
  - belongs to the class NP, and
  - is NP-hard, i.e.
    - > every problem in NP is reducible to circuit-satisfiability problem!
- Circuit-satisfiability problem
  - we say that a one-output Boolean combinational circuit is satisfiable
    - > if it has a satisfying assignment,
    - > a truth assignment (a set of Boolean input values) that causes the output of the circuit to be 1
- Proof...

## NP-Completeness Proofs

- Once we proved a NP-complete problem
- To show that the problem Q is NP-complete,
  - choose a know NP-complete problem P
  - reduce P to Q
- The logic is as follows:
  - since P is NP-complete,
    - > all problems R in NP are reducible to P,  $R \leq_p P$ .
  - show  $P \leq_p Q$
  - then all problem R in NP satisfy  $R \leq_p Q$ ,
    - > by transitivity of reductions
  - therefore Q is NP-complete

## Solving hard problems: Approximation Algorithms

- an algorithm that returns near-optimal solutions
- may use heuristic methods
  - > e.g. greedy heuristics
- Definition: Approximation algorithm
  - An approximation algorithm for a problem is
  - a polynomial-time algorithm that,
  - when given input I, outputs an element of FS(I).
- Definition: Feasible solution set
  - A feasible solution is
    - an object of the right type but not necessarily an optimal one.
  - FS(I) is the set of feasible solutions for I.

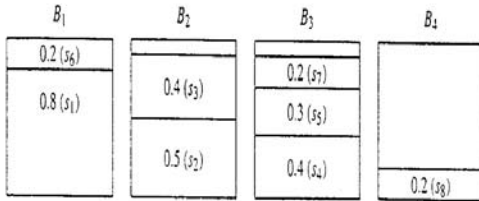
## Approximation Algorithm e.g. Bin Packing

- How to pack or store objects of various sizes and shapes
- with a minimum of wasted space
- Bin Packing
  - Let  $S = (s_1, \dots, s_n)$ 
    - > where  $0 < s_i \leq 1$  for  $1 \leq i \leq n$
  - pack  $s_1, \dots, s_n$  into as few bin as possible
    - > where each bin has capacity one
- Optimal solution for Bin Packing
  - considering all ways to
  - partition S into n or fewer subsets
  - there are more than
  - $(n/2)^{n/2}$  possible partitions

## Bin Packing: First fit decreasing strategy

- places an object in the first bin in which it fits
- $W(n)$  in  $\Theta(n^2)$

$S = (0.8, 0.5, 0.4, 0.4, 0.3, 0.2, 0.2, 0.2)$

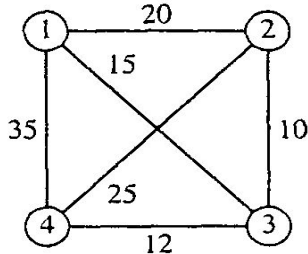


## Algorithm: Bin Packing (first fit decreasing)

- Input: A sequence  $S=(s_1, \dots, s_n)$  of type float, where  $0 < s_i < 1$  for  $1 \leq i \leq n$ .  $S$  represents the sizes of objects  $\{1, \dots, n\}$  to be placed in bins of capacity 1.0 each.
- Output: An array  $\text{bin}$  where for  $1 \leq i \leq n$ ,  $\text{bin}[i]$  is the number of the bin into which object  $i$  is placed. For simplicity, objects are indexed after being sorted in the algorithm. The array is passed in and the algorithm fills it.
- **binpackFFd(S,n,bin)**
- `float[] used=new float[n+1];`
- `//used[j] is the amount of space in bin j already used up.`
- `int i,j;`
- Initialize all used entries to 0.0
- Sort  $S$  into descending (nonincreasing) order, giving the sequence  $s_1 \geq s_2 \geq \dots \geq s_n$ .
- `for(i=1; i<=n; i++)`
- `//Look for a bin in which s[i] fits.`
- `for(j=1; j<=n; j++)`
- `if(used[j]+s_i<=1.0)`
- `bin[i]=j;`
- `used[j] += s_i;`
- `break; //exit for(j)`
- `//continue for(i).`

## The Traveling Salesperson Problem

- given a complete, weighted graph
- find a tour (a cycle through all the vertices) of minimum weight
- e.g.



## Approximation algorithm for TSP

- The Nearest-Neighbor Strategy
  - as in Prim's algorithm ...
- NearestTSP(V, E, W)
  - Select an arbitrary vertex  $s$  to start the cycle  $C$ .
  - $v = s$ ;
  - While there are vertices not yet in  $C$ :
    - Select an edge  $vw$  of minimum weight, where  $w$  is not in  $C$ .
    - Add edge  $vw$  to  $C$ ;
    - $v = w$ ;
  - Add the edge  $vs$  to  $C$ .
  - return  $C$ ;
- $W(n)$  in  $O(n^2)$ 
  - where  $n$  is the number of vertices