

Approximation Algorithms

CSCI 3100

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Introduction

In general, computer cannot solve NPC problem efficiently

But, many NPC problems are too important to abandon

If a problem is an NPC problem, you may try to

- find a pseudo polynomial time algorithm if it is not NPC in the strong sense
 - Pseudo-polynomial: polynomial in the numeric value of the input
 - Subset sum problem: $O(N \cdot K)$
- solve restricted problems
- find approximation algorithms
- Use heuristic based methods

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Approximation Algorithms

Let's consider optimization problems only.

An algorithm A is an approximation for a problem L : if given any valid instance I , it finds a solution $A(I)$ for L and $A(I)$ is "close" to optimal solution $OPT(I)$.

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Approximation Ratio

Approximation ratio (or bound) β of approximation algorithm A for problem L .

$A(I)/OPT(I) \leq \beta$; if L is a minimization problem and $A(I) \geq OPT(I) > 0$

$OPT(I)/A(I) \leq \beta$; if L is a maximization problem and $OPT(I) \geq A(I) > 0$

Example:

- Suppose we have an "approximation algorithm for the max-clique" problem.
- Given a graph G , the algorithm finds a clique of size M .
- If our approximation algorithm has approximation ratio β , we know that the optimal solution is no bigger than $\beta * M$
- Suppose β is 2 and M is 10.
- Then we know that the optimal solution is no bigger than 20.

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Maximum Programs Stored (PS) Problem

Optimization PS Problem: Given a set of n program and two storage devices. Let s_i be the amount of storage needed to store the i^{th} program. Let L be the storage capacity of each disk. Determine the maximum number of these n programs that can be stores on the two disks (without splitting a program over the disks).

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Example : $L = 10$ $S = (2, 4, 5, 6)$
How many programs (from the set S)
can we store on two disks of size 10?

- A. 1
- B. 2
- C. 3
- D. 4

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The decision PS problem is NPC

Approximation PS Algorithm

// assume programs are sorted in nondecreasing order of program size,

// i.e. $s_1 \leq s_2 \leq \dots \leq s_n$.

// Put as many programs as you can in the 1st disk, then go to the 2nd disk.

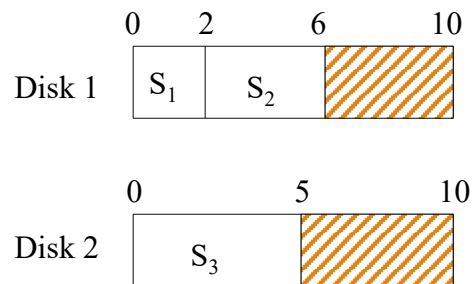
$i = 0$; $c=0$; // c count the number of stored program

```
for j = 1 to 2 {  
    sum = 0  
    while ( sum +  $s_i \leq L$  ) {  
        store  $i^{\text{th}}$  program into  $j^{\text{th}}$  device  
        sum +=  $s_i$   
        i++; c++  
        if i > n return  
    }  
}
```

Run time:

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Example : $L = 10$ $S_i = (2, 4, 5, 6)$



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Approximation Ratio

Let C^* be the optimal (maximum) number of programs that can be stores on the two disks.

The above approximation PS algorithm gives very good approximation ratio.

$$C^* \leq (C + 1) \quad \text{OR} \quad C^*/C \leq 1 + 1/C$$

i.e. the given program stores

at most 1 program less than
the optimal solution

The above approximation algorithm returns value C such that $C^* \leq C+1$, where C^* is the optimal solution.

Our example showed that there exists a set S and value L such that $C^* = C+1$.

Need to show that $\forall \{s_1, s_2, \dots, s_n\}$ and L , $C^* \leq (C + 1)$

Let's consider only one disk with capacity $2L$.

Greedy proof: We can store maximum number of programs into the disk by considering programs in the order of

$$s_1 \leq s_2 \leq \dots \leq s_n$$

Let α be the maximum number of programs that are stored in the disk

Clearly $\alpha \geq C^*$ and $s_1 + s_2 + \dots + s_\alpha \leq 2L$ (i)

Let j be an index such that

$$\begin{aligned} (s_1 + s_2 + \dots + s_j) &\leq L \text{ and} \\ (s_1 + s_2 + \dots + s_{j+1}) &> L \end{aligned} \quad (\text{ii})$$

$j \leq \alpha$ and j programs are stored in the 1st disk by the above approximation algorithm

By (i) & (ii), $(s_{j+2} + s_{j+3} + \dots + s_\alpha) \leq L$

→ $(s_{j+1} + s_{j+2} + \dots + s_{\alpha-1}) \leq L$
→ at least $(j+1)^{\text{th}}$ program, $(j+2)^{\text{th}}$ program, ..., $(\alpha-1)^{\text{th}}$ program are stored in 2nd disk by the above approximation alg. **Done!**

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Travelling Salesman Problem

Given

- a set of cities/clients and travel duration between cities/clients

Find

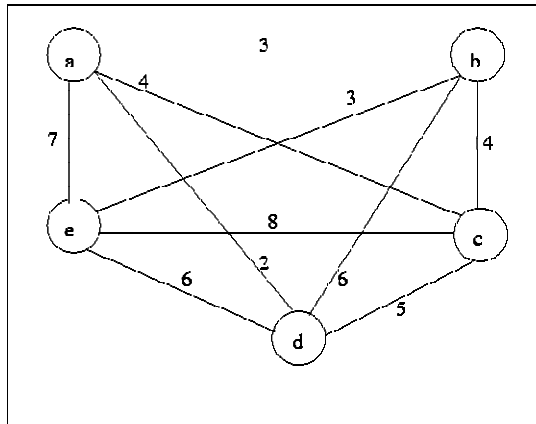
- A route that visits all cities/clients as quickly as possible, returning to the starting point

Translate this to a graph problem:

- Given a graph $G = (V, E)$, where V is a set of cities/clients and E is a weighted set of edges/roads connecting cities/clients and a vertex s in V ,
- Find a cycle that starts at s and visits each vertex such that the edge weights of this cycle are minimized
- Edges can be repeated

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Example: start at vertex a



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Approximation Algorithm

Let s be the starting point

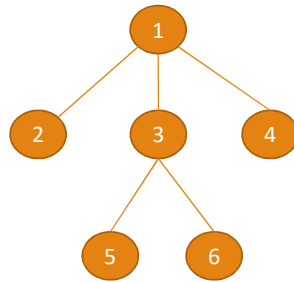
Construct MST starting at s

List vertices of the MST in the order visited by “pre-order” walk:

```
◦ pre-order-walk(T, root)
  visit root
  for (each vertex v adjacent to root)
    pre-order-walk(Subtree of T starting at v, v)
  visit v
```

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Pre-order walk starting at 1



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The above approximation algorithm for TSP has an approximation factor 2

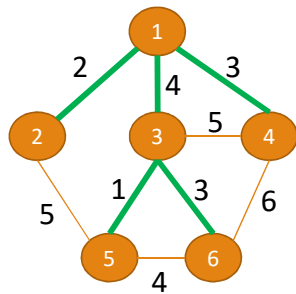
Let O be the cost of the optimal solution

We know that $O \geq \text{weight (MST)}$. Which of the following statements is true?

- A. $\text{weight (path produced by algorithm)} \leq 2 * \text{weight (MST)} \leq 2 * O$
- B. $O < 2 * \text{weight(MST)}$
- C. $\text{weight(path produced by algorithm)} < O$
- D. $2 * O < \text{weight(path produced by algorithm)}$
- E. None of the above

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Example: starting at vertex 1,
what is the weight of the
solution produced by our TSP
approximation algorithm.



A. 26

B. 13

C. 5

D. 20