# Advanced Algorithms - Homework 2

#### Question 1

First, we need to find the biggest size of vertices forming a maximal independent set (MIS), note this size by  $\mathbf{k}$ .

In order to find k we can perform a binary search in the range [1,n] (n = |V|) using the function  $A_d$  that runs in polynomial time.

 $\mbox{Note}$  - The first TRUE that  $A_d$  returns is not enough, we need to keep searching for the biggest.

**Runtime** to find the best k:  $log_2(n) \cdot T(A_d)$ 

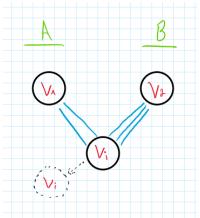
### $A_0$ :

After we know the value of k we can start the second part of the solution, run n iterations and each iteration we'll delete a vertex and check with  $A_d(G',k)$  if there is still MIS by the size of k and if the graph has k left vertices we can quit the loop and return the graph, if there is not MIS by the size of k (the decision function returns FALSE) then we need to undo the deletion of the vertex and continue to the next iteration.

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\label{eq:Pseudo:} \begin{split} & A_o(G) \\ * \ check \ edge \ cases \\ & k = binary\_search(G) \ \# \ as \ described \ above \\ & for \ v \ in \ V: \\ & G. delete\_vertex(v) \\ & if \ A_d(G,k): \\ & if \ |V| = k: \\ & return \ vertices \ \& \ edges \\ & else: \\ & G.add\_vertex(v) \end{split}
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Runtime of  $A_o(G)$ :  $|V| \cdot T(A_d) = n \cdot polynomial$ -time

First, we need to understand how the algorithm works, lets see the example below



As we can see  $c(v_i, A) = 2$  and  $c(v_i, B) = 3$ 

The algorithm works in a greedy way  $\Rightarrow$  in this example the algorithm will decide to assign  $v_i$  to group A. We can see that the worst scenario is if  $c(v_i,A)=c(v_i,B)$ , therefore half of the edges will be in the maximum cut and half will be "lost" (r=2) Note - the worst case for this algorithm is when the equality I just mentioned above repeats itself for each of the vertices and their edges equal distributed to  $v_1, v_2$  (A, B respectively) without edge between  $v_1$  and  $v_2$ 

| Example for the worst case                                  | The algorithm will produce maximum cut size of <b>2</b> but the optimal solution is <b>4</b> |
|---|--|
| V <sub>2</sub> V <sub>3</sub> V <sub>4</sub> V <sub>4</sub> | V <sub>A</sub> V <sub>3</sub> V <sub>4</sub>   |

We can see the simplest example above.

Let G be a **complete** graph with weights that obey the **triangle inequality**, and a subset of vertices R (terminals).

Let OPT be the cost of an optimal solution  $T^*$  to Steiner tree problem.

We'll start by doubling each edge to obtain an Euler cycle, with DFS tour we know this cost is  $2 \cdot OPT$ .

Making a Hamilton circuit using "short-cutting" Steiner vertices and visited vertices (shortcut = connecting new edge between pair of adjacent terminals in the DFS).

The short-cutting doesn't increase the total of the cost (complete graph & triangle inequality). Deleting one edge from this Hamilton circuit yields a spanning tree of R with cost at most  $2 \cdot OPT$ .

So, if this tree is a MST on R we know its cost is less than  $2 \cdot OPT$ 

#### To summary:

 $2 \cdot OPT = cost$  of the tour on the terminals in  $T^*$ 

≥ cost of any spanning tree (the tree after the manipulations, adding edges short-cutting and so on)

≥ cost of the MST of the tree

Let b be the number of the bins after using the algo Next-fit and OPT be the number of bins possible that needed to pack all of the items.

Two things we know:

- 1. The maximum size of each item is  $\frac{1}{5}$
- 2. Each of the bins full at least  $\frac{4}{5}$  of the maximum capacity (max cap. = 1) (from 1.)
- 3.  $bin_{last} + bin_{last-1} > 1$

The total size of the items is at least  $\frac{4}{5} \cdot (b-2) + 1 \rightarrow OPT > \frac{4}{5} \cdot (b-2) + 1 = \frac{4b}{5} \cdot \frac{3}{5}$ Order the equation will results us that  $\frac{5}{4}OPT + \frac{3}{4} > b$ 

Let s be a sequence as the following  $s = (\frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \varepsilon)$  where  $\varepsilon \leq \frac{4}{5n}$ .

Consider our instance as  $\overline{s, s, \cdots, s, s}$ ,  $\frac{1}{5}$  therefore we have  $4 \cdot n + 1$  of the items  $\frac{1}{5}$  and we have n of the items  $\varepsilon$ .

The algo Next-fit will pack this instance using n+1 bins and the OPT will do with  $\frac{4n}{5}+1$  bins.

$$\frac{5}{4}$$
 · OPT- $\frac{1}{4} = \frac{5}{4} \left(\frac{4n}{5} + 1\right) - \frac{1}{4} = n + 1$  exactly as Next-fit algo.

**Note** -  $\varepsilon$  derive from the last bin (in optimal solution),  $\frac{1-\frac{1}{5}}{n} = \frac{\left(\frac{4}{5}\right)}{n} = \frac{4}{5n}$ 

I read and understood the example and the tightness of the analysis.