

# ASSIGNMENT 7

ALGORITHMS & COMPLEXITY (CIS 522-01)

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# 1. File transfer

## Problem Model

In this problem we will have  $k$  network sources from source to destination, and  $n$  files that we want to transfer from source to destination. Each one of the files  $f_i$  has a length  $l_i$ . We want to find out how to assign the files to each one of the  $k$  routes so we can transfer the  $n$  files in the shortest time.

We will also assume in this problem that the network capacity is unlimited for each one of the networks.

## Class

This algorithm belongs to the P class, as it can be implemented as an approximation greedy algorithm for load balancing.

## Algorithm

In this algorithm, we will first sort the file lengths in decreasing order, and then we will start scheduling each one of the files to the less loaded network at each step. We will terminate this execution whenever we schedule all the files.

## Pseudocode

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**Algorithm 1** File transfers pseudocode

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- 1: We first sort all the files  $f_i$  in decreasing order
  - 2: **while** We didn't assign all the files **do**
  - 3:     Assign file  $f_i$  to the least loaded network  $N_i$
  - 4: **end while**
  - 5: **return** The list of the list assignments
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## Problem instance

An example of how to solve this problem can be found in the image attached to this assignment *Problem1.jpg*.

## Time complexity

The time complexity of this implementation is  $O(n \log m)$ , as when we want to schedule each one of the jobs, we have to find which is the machine, or in this case the network, with the least load.

## Approximation guarantee

The approximation guarantee for load balancing with list-scheduling is:

$$L \leq \frac{3}{2}L^*$$

## 2. Multiple Interval Scheduling

### Problem Model

In this problem we have a processor that can run jobs over some period of time. People submit jobs, and the processor can only work on one job at a time. These jobs run in a set of intervals of time, so if you accept that job, you have to accept all the job intervals.

We are given  $k$  jobs and we want to find out if we can schedule all  $k$  jobs in such a way that not two jobs of the accepted jobs overlap.

We will model this problem as a *Interval coloring* problem, in which we have a set of intervals, and we have to schedule all them in a way that we don't use more than the number of colors that we have available, in this case the number of processors, one. Also, in this problem if we assign one wave to one color we have to continue assigning this waves to the same color, same as here that if we schedule one job, we have to schedule all the intervals corresponding to this job.

### Class

The *Graph coloring* problem can be broken down into *Interval coloring* problem, so it belongs to NP-Complete problems.

## Complexity of the problem

We will follow the following strategy to prove that our problem is NP-Complete.

- Prove that  $X \in NP$
- Choose a problem  $Y$  that is known to be NP-Complete
- Prove that  $Y \leq_p X$

For our given problem, the certifier  $B$  will prove that all the jobs are scheduled in at least one process, and at the same time that not two jobs are scheduled at the same time.

Then we choose a problem  $Y$ , in this case the graph coloring problem, that we know that is NP-Complete.

In the case of graph coloring, we have  $k$  colors, and a Graph  $G$ , we want to find out if there is a possible  $k$ -coloring for that graph. There are also conflicts between objects, which means that conflicting objects can't go into the same set. Also, the coloring in the graph must be consistent, so if we assign a color to one wavelength, the next time this wavelength appears, we must assign these wavelength the same color as before.

In the case of our problem, we have  $n$  jobs, and one processor, so the depth of our graph is only one. We also want to see if we can schedule  $k$  of the jobs, while taking into account that when we schedule a job we have to schedule all the intervals related to this job, and that not two jobs can run at the same time in the same processor.

So what we will do is to solve this problem as if it was an interval scheduling problem, having unlimited colors, or unlimited depth then we will check how many jobs are scheduled in one of the colors. If the number of jobs scheduled in this color is equal or greater than  $k$  that means that the answer to our question will be yes, otherwise we will need more processors to schedule those  $k$  jobs.

## 3. Airport service

### Problem Model

In this problem we have  $n$  airport sites, and  $m$  direct flights scheduled between these airports. Building a service facility that allow us to operate in some airport

$A_i$  has a cost  $C_i$ . We want to select a subset of airports in such a way that we can have all airports connected, at the minimum cost possible.

We will model this problem as a directed graph  $G$ , in which the airports will be the nodes, and the flights will be the edges. Also, each one of this nodes will have a cost associated  $C_i$  that will be the cost of building the service facility in that airport  $A_i$ .

## Class

This problem belongs to class  $P$ , because it can be solved by an approximation method using a *greedy* approach.

## Algorithm

To solve this problem, we will use a greedy vertex cover approach, or as it is usually named, the *pricing method*.

In this algorithm, we will first set the price paid for every edge to 0, and then , we will continuously increase the edges values, untill one of the nodes at one of the sides of the edge is 'tight', at that point we will add that node to the set of tight nodes  $S$ . We will then continue to look for edges that are not connected to tight nodes, to increase the edge value until one of the nodes that the edge is connected to becomes tight. We will continue to do this until we can't find an edge which is not connected to any tight node. Then we will return the set of tight nodes, which will be the solution to our problem. In our case the nodes returned would be the airports in which we should build the service facilities.

## Pseudocode

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**Algorithm 2** Pricing method pseudocode

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- 1: We first set  $p_e = 0$  for all edges  $e \in E$
  - 2: **while** There is an edge  $e = (i, j)$  such that neither  $i$  nor  $j$  is tight **do**
  - 3:     We select an edge  $e$
  - 4:     We increase  $p_e$  without violating fairness
  - 5: **end while**
  - 6: **return**  $S$  which is the set of all tight nodes
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## Problem instance

A small problem instance of this algorithm can be found in *Figure 11.8* in Algorithm Design book page 622.

## Time complexity

The time complexity of this algorithm will be  $O(m)$ , where  $m$  is the number of edges, or number of direct flights. As in the worst case scenario, we will have to go through all edges to find all the tight nodes.

## Approximation guarantee

The approximation guarantee for this algorithm is:

$$w(S) \leq 2 \sum_{e \in E} p_e \leq 2w(S^*)$$

# 4. Telecommunication company

## Problem Model

In this problem, we have to build base stations in various locations to provide mobile phone service. We have houses located in  $n$  different locations. We want to select  $k$  locations to build  $k$  base stations so that the maximum distance from any house to its closest station is minimized.

## Class

This algorithm belongs to  $P$  class, as it can be implemented as an approximation algorithm with a greedy approach.

## Algorithm

In this algorithm, we will start by setting one of the houses in the  $n$  locations to build the base station, then we will build the next location on the house that

is further from the set of centers or base stations. We will continue doing this until we can't place any more base stations.

## Pseudocode

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**Algorithm 3** Center selection pseudocode

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1: We assume that the number of centers is fewer than the number of points
2: Otherwise, our centers will be placed in top of our locations
3: We first select any site  $s$  and  $C = s$ 
4: while  $|C| < k$  do
5:   Select a site  $s \in S$  that maximizes  $dist(s, C)$ 
6:   Add site  $s$  to  $C$ 
7: end while
8: return  $C$  as the selected set of sites
```

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## Problem instance

An example of how to solve this problem can be found in the image attached to this assignment *Problem4.jpg*.

## Time complexity

The time complexity of this algorithm will be  $O(k)$ , where  $k$  is the number of points that we are looking to place.

## Approximation guarantee

This greedy algorithm returns a set  $C$  of  $k$  points such that  $r(C) \leq 2r(C^*)$ , where  $C^*$  is an optimal set of  $k$  points.

## 5. Job requests

### Problem Model

In this problem, we have  $n$  job requests to select from, each job paying  $p_i$  amount of reward. Some of these jobs can't be scheduled together due to time conflicts.

Our goal is to select the set of jobs that have no conflict among themselves that gives us the highest total amount of reward.

We will model this problem as a tree, in which the nodes are the jobs, and the edges are the conflicts. Each one of the nodes (jobs) will have associated a certain amount, that will be the reward for completing that job. Then our goal will be to find the maximum-weight independent set on the tree.

## Class

The problem of finding a maximum weight independent set in a tree belongs to  $P$  class.

## Algorithm

We will first root an arbitrary node  $r$ , to orientate all the tree's edges away from  $r$ . For a node  $u$  the nodes that are in between this node and  $r$  will be then the parent nodes, and the nodes that face the other way will be the children nodes.

Then we will use a dynamic programming solution at each node, in which we will have to choose to include that node or not depending on which gives the maximum revenue, if including that node, but not its children, or including the children nodes but not that node.

## Pseudocode

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**Algorithm 4** Find maximum-weight independent set pseudocode

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1: First root the tree at arbitrary node  $r$ 
2: for All nodes  $u \in T$  in post-order do
3:   if  $u$  it's a leaf node then
4:      $M_{out}[u] = 0$ 
5:      $M_{in}[u] = w_u$ 
6:   else  $u$  it's not a leaf node
7:      $M_{out}[u] = \sum_{v \in children(u)} \max(M_{out}[v], M_{in}[v])$ 
8:      $M_{in}[u] = w_u + \sum_{v \in children(u)} M_{out}[v]$ 
9:   end if
10: end for return  $\max(M_{out}[r], M_{in}[r])$ 

```

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### **Problem instance**

An example of how to solve this problem can be found in the image attached to this assignment *Problem5.jpg*.

### **Time complexity**

The time complexity of this algorithm will be  $O(n)$ , where  $n$  is the number of nodes, in this case the number of jobs.

## **6. City roads monitoring**

### **Problem Model**

We will model this problem as a graph  $G$ , in which the nodes will be the cross-roads, and the edges will be the roads.

### **Class**

As  $k$  is small in this problem, we can solve this vertex cover problem in polynomial time. Then this problem belongs to P.

### **Algorithm**

Our algorithm will check all the possible subsets of  $V$  of size  $k$ , and check if any of them is a vertex cover. This algorithm implementation will be based on brute force.

## Pseudocode

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**Algorithm 5** Vertex cover pseudocode

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1: while There is a subset  $S$  of  $V$  of size  $k$  that we didn't check yet do
2:   We check if subset  $S$  is a vertex cover in our graph
3:   if It's a vertex cover then
4:     return Vertex cover  $S$ 
5:   Break the loop
6:   else if It's not a vertex cover then
7:     We continue with the next possible subset
8:   end if
9: end while
```

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## Problem instance

An example of how to solve this problem can be found in the image attached to this assignment *Problem6.jpg*.

## Time complexity

There are  $\binom{n}{k}$  possible subsets, and as it takes  $O(kn)$  time to check whether a subset is a vertex cover or not, the time complexity of our algorithm will be  $O(kn^{\binom{n}{k}+1})$