

# CS5800: Algorithms Spring 2018

## Assignment 4.1

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February 23, 2018

### (a) Algorithm:

Let `benefit` is a 2D array holding benefit of each project for all possible number of programmers

1. Set `assignedProgrammers`, `maxProgrammersAssignedToAProject` and `projVal` to 0
2. Create empty array `numProgsAssigned` of size = number of projects
3. Till `assignedProgrammers < number of programmers`, repeat following steps:
  - a. Set `maxPossibleBenefitFromAnyProject` to integer min value
  - b. Set `maxPossibleProjIndex` to -1
  - c. Set `i=0`. For each project, do the following steps:
    - i. set `benefit` to 0
    - ii. If project doesn't have any programmer assigned  
Set `benefit` to benefit produced by assigning single programmer i.e.  
`benefitArr[i][ numProgsAssigned[i] ]`  
Else  
Set `benefit` to benefit produced by new programmer i.e.  
`benefitArr[i][ numProgsAssigned[i] ] - benefitArr[i][lastProgrammerIndex]`
    - iii. If `benefit` is more than `maxPossibleBenefitFromAnyProject`  
set `benefit` to `maxPossibleBenefitFromAnyProject`  
set `i` to `maxPossibleProjIndex`
  - d. Increment `numProgsAssigned[maxPossibleProjIndex]` by 1
  - e. Add `maxPossibleBenefitFromAnyProject` to `projVal`
4. Return `projVal` to print max benefit of all projects  
or `numProgsAssigned` to print actual project assignments

(b) **Correctness:** To prove the greedy algorithm, we have to prove following properties:

**1. Optimal substructure:** Let  $A_{ij}$  be the optimal solution or max benefit for projects  $i$  to  $j$ . Say, it includes a solution  $a_k$  which is benefit from  $k^{th}$  project. So,  $A_{ij}$  can be written as  $A_{ik} + A_{kj} + a_k$ . Now  $A_{ij}$  should contain optimal solution of its subproblems. If its not true, then there exists a solution  $A'_{ik} > A_{ik}$ . So, if we add  $a_k$  to the solution (benefit of  $k$ th project, which is not there in  $A_{ik}$ ) we get a new solution  $A'_{ik} + a_k + A_{kj} > A_{ik} + A_{kj} + a_k$ . So,  $A'_{ij} > A_{ij}$ , which is contradiction to original solution.

So, the problem exhibits optimal substructure property i.e. its optimal solution contains optimal solution of subproblems.

**2. Greedy choice:** To have a greedy choice for this solution means to have as much benefit possible from a project. If we make a greedy choice, we are left with one subproblem i.e. getting benefits from other projects. For the first choice we don't need to consider other projects because we are always choosing the project with maximum benefit.

Now to prove greedy choice is optimal, let us assume  $a_m$  is the project with max benefit. Let  $A_k$  be the optimal solution where  $a_j$  is the project where first programmer is assigned. Now, if  $a_j \neq a_m$ , then we can have a new solution  $A'_k = A_k - a_j + a_m$ . Since  $a_m$  gives max benefit, we can write  $a_m > a_j$ . So,  $A'_k > A_k$ , which is a contradiction to original solution. Hence,  $a_j = a_m$  i.e. the greedy choice is always part of the optimal solution.

### Complexity analysis:

Here we are calculating benefit of each project for every programmer. So, for  $m$  projects and  $n$  programmers, it should

take  $O(mn)$  time. Time complexity =  $O(mn)$ . The amount of space taken by the algorithm changes depending number of projects. numProgsAssigned is the array that holds project assignments and is of size number of projects. So for  $m$  projects, space complexity should  $O(m)$ .