Algorithm Design - HW1

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1 Exercise 1

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Algorithm 1: Help Philip

Result: Find optimal expected reward

Initialize N, K, costlist[K], M[N+1][K];

for box \leftarrow K to-1 do

| for rew \leftarrow 0 to N+1 do

| t = expected(box,rew);
| M[rew][box] = max(t-c[box],rew);
| end

end

return M[0][0];
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Creates a table $K \cdot (N+1)$ with all possible expected wins. We start from last box cause is the one without constraints on next box. t represent the expected value given the following boxes (thats the reason why we start from last one). Expected value at box b and reward rew is calculated as following:

$$t = \frac{M[rew][b+1]*(rew+1)}{N+1} + \sum_{i=rew+1}^{N} \frac{M[i][b+1]}{N+1}$$

Note: on last box replace M[rew][b+1]*(rew+1) with rew and M[i][b+1] with i to avoid out-of-bound exception.

The result we want in return is the first cell because it represent our initial state (box = 0, rew = 0). The cost is $O(n^2 \cdot k)$.

2 Exercise 2

Make use of UNION - FIND structures like in Kruskal's algorithm:

FIND - SET(v) returns the set containing node v;

UNION(s1, s2) unify the two sets s1 and s2 in a new set of size |s1| + |s2|.

Sort edges in crescent order according to their weight. T = total weight of complete graph. For every edge e = (v1,v2) (from min to max weight) compute FIND-SET(v1) = X and FIND-SET(v2) = Y and then UNION(X,Y) to add both to the "wannabe" complete graph. We want a complete graph so there must be an edge connecting each pair of node: for this reason, each time we do an union of sets, we should add new edges connecting each node of the two sets X and Y. These edges must be heavier than e(weight(e) + 1) so that the Kruskal's algoritm will avoid to add them in the MST. To calculate the total weight of the graph, at each step:

$$T = T + weight(e) + (|X| \cdot |Y| - 1) \cdot (weight(e) + 1)$$

Creating UNION-FIND structures and ordering the edges in ascending order will cost $O(|V| \cdot \log |V|)$ with a simple merge-sort, iterating over all edges will cost O(|V|), so the total cost will be $O(|V| \cdot \log |V|)$ to find the total minimal weight of the graph.

If we want to compute even the complete graph itself, each time we make the UNION operation, we have to add the edges to the graph: this will cost $|V|^2$ due to the fact that for each vertex in X we iterate over all vertices in Y.

3 Exercise 3

The goal is to compute how much chocolate Federico can sell according to his friends network: for this purpose we model a network flow as follow.

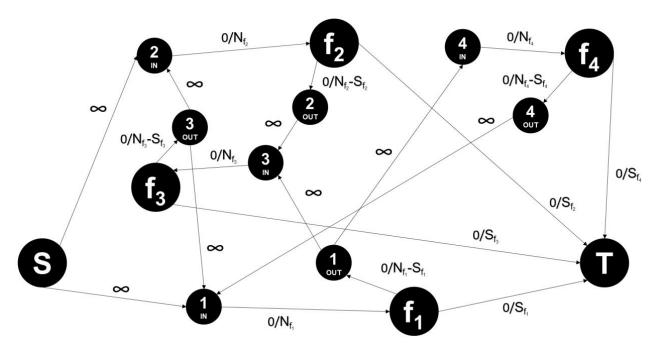
Of course the SOURCE (S) will be Federico with available resources $= \infty$. We can imagine the SINK (T) instead as the total of possible customers and therefore the MAX-FLOW of the graph will give us the maximum of chocolate that Federico can sell in a particular week. F is the set of vertices representing Federico's friends.

Every vertex $f_i \subseteq F$ is connected with an IN_i vertex with an incoming edge of capacity N_{f_i} representing the max storage capacity of f_i .

Every vertex $f_i \subseteq F$ is connected with an OUT_i vertex with an outgoing edge of capacity $N_{f_i} - S_{f_i}$ representing the max quantity of chocolate that f_i can exchange with other friends.

Every vertex $f_i \subseteq F$ is connected with T with and outgoing edge of capacity S_{f_i} representing the max quantity of chocolate that f_i can sell to customers.

Federico knows every connection between his friends and each other and of course with him: lets call the subset of friends that he meets regularly X. Federico is connected to every $f \subseteq X$ with an edge of capacity ∞ to the corresponding IN_i . Set X represents the only way that Federico can use to put the chocolate into the network, the only direct connection between him and customers. Every $f \subseteq F$ could meet each other and exchange exceeding chocolate $N_f - S_f$: connect every pair of "meeting friends" i, j with two edges $e_1 = (OUT_i, IN_j)$ and $e_2 = (OUT_j, IN_i)$ of capacity equal to ∞ . Example following.



In the new semester every $f \subseteq F$ will have an associated university building $b_f \subseteq B$ with a certain number of customers available c_b . To model this scenario we can simply add edges from every f_i to their corresponding b_i with capacity S_{f_i} and then from every b_i to the sink with capacity c_{b_i} . Federico's business will be lowered only if for every b_i the sum of capacity of incoming edges is > than the outgoing edge capacity.

4 Exercise 4

The goal of the exercise is to demonstrate that the problem is NP-complete: we will do this demonstrating that it is NP and NP-hard.

NP demonstration A problem is NP if given an instance of it and a possible solution for it, it requires a poly-time algorithm to tell if it is valid or not. In our case a possible solution is a complete sequence of jobs that Giovanni has to do. To check it we just need to iterate over it and verify, for each job j_i , that, given its starting time t_i :

- $t_i \ge s_i$ to avoid a job starting before its earliest time.
- $t_i + l_i \leq d_i$ to avoid job finishing after its deadline.
- $t_i + l_i \le t_{i+1}$ to avoid overlapping with next job.

Iteration requires O(N) time \to the problem is NP.

NP-hard demonstration We will do this reducing Subset-Sum, a well known NP-complete problem, to this problem. Suppose we have n numbers w_i and our goal is to find a subset of sum W. We now create a similar version for scheduling: $K = \sum_{i=1}^{n} w_i$, n jobs j_i each with $s_i = 0$ and $d_i = S + 1$ (same earliest time and deadline) and $l_i = w_i$. In this way we can set them in any order, all finishing in time. Now lets add one more constrain: we want to be able to solve it only by grouping together a subset of jobs whose durations sum up to W. For doing this we define job j_{n+1} with $s_{n+1} = W$, $d_{n+1} = W + 1$ and $l_{n+1} = 1$. In a possible solution, job j_{n+1} can be run only in [W, W + 1] leaving K time between $s_i = 0$ and $d_i = K + 1$: in this way every job must be executed one by one without pauses. Assuming that jobs j_i with $i = 1 \dots n$ are the ones that run before time W than $w_{i_1} \dots w_{i_n}$ is the Subset that sum up to W. If we have $w_{i_1} \dots w_{i_n}$ we can schedule them before j_{n+1} and the remainder after j_{n+1} obtaining a feasible solution.