

CSC373    Fall'20  
Final Assessment Solutions  
Date: December 18, 2020

**Q1 [10 Points] Find the Single Attendee**

There are  $2n + 1$  attendees at a party, which includes  $n$  couples and a single person. At the end of the party, all the attendees form a line in which each person stands next to their partner, except for the single person, who stands somewhere in the line. As an example, for  $n = 3$ , the seven attendees could be standing in the order  $(A_1, A_2, B_1, B_2, C, D_1, D_2)$ , where  $(A_1, A_2)$ ,  $(B_1, B_2)$ , and  $(D_1, D_2)$  are couples and  $C$  is single.

Your job is to find the position of the single person (this would be 5 in the above example). But you don't know which ones are partners. All you can do is ask questions of the form "Are the  $i$ -th and  $j$ -th people in the line partners?" Design a divide-and-conquer algorithm for this problem which finds the position of the single person by asking  $O(\log n)$  questions. Justify your answer.

**Solution to Q1**

Suppose  $A$  is the array of attendees. The key idea is to compare two attendees about half-way in the array. Suppose we compare  $A[n]$  and  $A[n + 1]$  and they are a couple. If  $n$  is odd, then  $A[1 \dots n + 1]$  is of even length, which means it must contain  $(n + 1)/2$  couples and not the singleton. So we can search  $A[n + 2 \dots 2n + 1]$  for the singleton. Similarly, if  $n$  is even, then  $A[1 \dots n + 1]$  is of odd length, which means it must contain  $n/2$  couples and the singleton. So we can search  $A[1 \dots n - 1]$  for the singleton (since we already know that  $A[n]$  and  $A[n + 1]$  are not the singleton). Similar conclusions hold if  $A[n]$  and  $A[n + 1]$  are not a couple. Also, note that we are careful to always call our algorithm on an array of odd length that contains the singleton.

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**Algorithm 1:** Find-the-Singleton

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Input: Array  $A$  of length  $2n + 1$ 
1 if  $n = 0$  then
2   | return 1
3 end
4 if If  $A[n]$  and  $A[n + 1]$  are a couple then
5   | // Search the second half if  $n$  is odd and the first half if  $n$  is even.
6   | return  $(n + 1) + \text{Find-the-Singleton}(A[n + 2 \dots 2n + 1])$  if  $n$  is odd and
7   |    $\text{Find-the-Singleton}(A[1 \dots n - 1])$  if  $n$  is even
8 else
9   | // Search the second half if  $n$  is even and the first half if  $n$  is odd.
10  | return  $n + \text{Find-the-Singleton}(A[n + 1 \dots 2n + 1])$  if  $n$  is even and
11  |    $\text{Find-the-Singleton}(A[1 \dots n])$  if  $n$  is odd
12 end
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For the worst-case number of questions, note that solving a list of length  $2n + 1$  requires solving a list of length at most  $n + 1$  and a single additional question. Hence, we have the recurrence relation

$T(2n + 1) \leq T(n + 1) + 1$ , which, by the master theorem, gives us  $T(n) = O(\log n)$ .

## Q2 [15 Points] Event Planner

There are  $n$  events, each takes one unit of time. Each event  $i$  will provide a profit of  $g_i$  dollars if it is started at or before time  $t_i$ , but will provide zero profit if it is not started by time  $t_i$  (so there is no point in scheduling event  $i$  unless it can be scheduled to start by time  $t_i$ ). Here,  $g_i, t_i \geq 0$  and  $t_i$  may *NOT* be an integer. An event can start as early as time 0 and no two events can be running simultaneously. The goal is to feasibly schedule a subset of the events to maximize the total profit.

(a) [2.5 Points] Prove that there exists an optimal schedule  $OPT$  in which every event that is scheduled is scheduled to start at an integral time. Note that in such a solution, each event  $i$  is either scheduled to start by time  $\lfloor t_i \rfloor$  or not scheduled at all.

(b) [5 Points] Design an efficient greedy algorithm which only schedules events at integral start times. [Hint: Let  $T = \max_i \lfloor t_i \rfloor$ . Think about which event you would schedule to start at time  $T$ .]

(c) [5 Points] Prove that your algorithm always returns an optimal solution.

(d) [2.5 Points] Analyze the worst-case running time of your algorithm. Explicitly state the data structures that your algorithm uses.

## Solution to Q2

(a) Consider any optimal schedule  $OPT'$  which schedules a subset of the events  $S$  and each  $i \in S$  is scheduled to start at  $s'_i$ . Next, consider the schedule  $OPT$  which also schedules the same set of events  $S$  but schedules each  $i \in S$  to start at  $s_i = \lfloor s'_i \rfloor$ .

Since  $s_i \leq s'_i$  for each  $i \in S$ , we know that each event in  $S$  is still scheduled profitably. Thus,  $OPT$  has the same profit as  $OPT'$  and it only schedules events to start at integral times. It remains to show that no two events are overlapping in  $OPT$ . But since events start at integral times and run for one unit of time, this is equivalent to proving that no two events have the same starting time under  $OPT$ .

To see this, consider any two events  $i, j \in S$ . Since  $OPT'$  is feasible,  $[s'_i, s'_i + 1)$  and  $[s'_j, s'_j + 1)$  must not overlap. This directly implies that  $\lfloor s'_i \rfloor \neq \lfloor s'_j \rfloor$ , i.e.,  $s_i \neq s_j$ , as required.

(b) We sort the events by their start deadlines and then, in a single pass, divide them into blocks  $E_0, \dots, E_T$  such that for each  $k \in \{0, 1, \dots, T\}$ ,  $E_k = \{i : k \leq t_i < k + 1\}$ . Note that events in  $E_k$  are profitable if started at time  $k$  or earlier but not if started at time  $k + 1$  or later. And note that  $T$  is the latest time at which we can start an event profitably.

Only events in  $E_T$  can be scheduled at time  $T$ . Among them, we schedule the most profitable one at time  $T$ . Then, we consider the unscheduled events in  $E_T$  along with the events in  $E_{T-1}$ , and schedule the most profitable among them at time  $T - 1$ . We continue doing this until we reach time 0. This is explained in the algorithm below.

At time  $k$ , to find the most profitable event among the unscheduled events in  $E_{k+1}, \dots, E_T$  along with the events in  $E_k$ , we maintain a priority queue of events from which we can find the most

profitable one and delete the scheduled one quickly.

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**Algorithm 2:** Greedy-Event-Scheduling

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1 Sort the events so that  $t_1 \leq \dots \leq t_n$ 
2  $T \leftarrow \max_i \lfloor t_i \rfloor = \lfloor t_n \rfloor$ 
3 Divide the events into buckets  $E_0, \dots, E_T$  such that  $E_k = \{i : k \leq t_i < k + 1\}$  for each
    $k \in \{0, 1, \dots, T\}$ 
4  $Q \leftarrow$  empty priority queue
5 for  $t = T, T - 1, \dots, 0$  do
6   Add all events in  $E_t$  to  $Q$  with their profit as the key
7   if  $Q$  is empty then
8     continue
9   end
10   $i \leftarrow$  most profitable event in  $Q$ 
11  Schedule event  $i$  to start at time  $s_i = t$ 
12  Delete event  $i$  from  $Q$ 
13 end

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(c) We say that a schedule is integral if it only schedules events at integral start times. Part (a) shows that there exists an optimal schedule that is integral. We say that two integral schedules *match* at time  $t$  if either both schedule the same event at  $t$  or both do not schedule any event at  $t$ . The *match level* of two integral schedules is the smallest  $t$  for which they match at time  $t$ .

Let  $G$  denote our greedy schedule. Among all optimal integral schedules, let  $OPT$  be the one with the smallest match level with  $G$ . If this match level is 0, then the greedy schedule is optimal, so we are done. Otherwise, suppose this match level is  $t + 1$ . Consider time  $t$ . There are three possibilities:

1.  $OPT$  schedules nothing at time  $t$  while  $G$  schedules some event  $i$ . Then,  $i$  must be scheduled at some other time in  $OPT$ , otherwise scheduling  $i$  at time  $t$  would not cause any conflicts and increase the profit, which is impossible. Now, changing the start time of  $i$  to  $t$  produces an integral optimal schedule  $OPT'$  which has the same profit (event  $i$  remains profitable when started at time  $t$  because  $G$  schedules it at time  $t$ ) and has match level  $t$  with  $G$ , a contradiction.
2.  $OPT$  schedules some event  $i$  at time  $t$  while  $G$  schedules nothing. Since  $G$  and  $OPT$  match at times  $t + 1, \dots, T$ ,  $i$  must be unscheduled when the greedy algorithm reaches iteration for time  $t$ . Since  $i$  is profitable if scheduled at  $t$ ,  $G$  cannot schedule nothing at time  $t$ , a contradiction.
3.  $OPT$  schedules some event  $i$  at time  $t$  while  $G$  schedules a different event  $j$ . Since  $G$  and  $OPT$  match at times  $t + 1, \dots, T$ , neither have  $i$  or  $j$  scheduled after time  $t$ . Further, since  $i$  must be unscheduled at the iteration for time  $t$  in  $G$ , but it schedules event  $j$ , the profit of  $j$  must be at least as much as the profit of  $i$ . So if  $OPT$  doesn't schedule  $j$ , then we can replace  $i$  by  $j$  at time  $t$  in  $OPT$ , and if  $OPT$  does schedule  $j$  at an earlier time, then we can swap the starting times of  $i$  and  $j$ . Note that  $j$  still remains profitable since it has the same starting time as in  $OPT$ , and  $i$  only moves early so remains profitable as well. In either case, we have a new integral optimal schedule with match level  $t$  with  $G$ , a contradiction.

(d) Sorting and grouping the events by the floor of their start time (Lines 1-3) takes  $O(n \log n)$

time. Creating the buckets then takes  $O(n + T)$  time. The loop runs for  $O(T)$  iterations, and in each iteration, finding the most profitable event and deleting the scheduled event from the priority queue takes  $O(\log n)$  time. Hence, the total running time is  $O((n + T) \log n)$ , which is not polynomial in the input length because  $T$  can be quite large.

However, we can slightly modify the algorithm such that we only create and store non-empty buckets, and every time we have an empty  $Q$  in Line 7, we reduce  $t$  directly to the time of the next non-empty bucket. Thus, we can reduce the number of iterations to the order of the number of events scheduled, which is  $O(n)$ . This reduces the running time to  $O(n \log n)$ , which is polynomial. While the loop runs for  $O(T)$  steps as stated, we can slightly modify it to skip over all the consecutive trivial steps (i.e. steps in which  $Q$  is empty in Line 7), thus

### Q3 [15 Points] Protect the Paintings Again

Recall the question about protecting paintings from midterm 1. A corridor of a museum is represented by the interval  $[a, b]$  (with  $a < b$ ) and contains valuable paintings. There are  $n$  guards stationed along the corridor. Guard  $i$  can protect the interval  $[s_i, f_i]$ , where  $a \leq s_i \leq f_i \leq b$ . We say that a subset of guards  $P \subseteq \{1, \dots, n\}$  is *acceptable* if the guards in  $P$  already collectively protect the entire corridor, i.e.,  $\cup_{i \in P} [s_i, f_i] = [a, b]$ . Assume that the set of all guards  $\{1, \dots, n\}$  is acceptable, so there is at least one acceptable set.

In the midterm, we designed a greedy algorithm for finding an acceptable subset  $P$  of *minimum cardinality*  $|P|$ . Instead, suppose that each guard  $i$  has an associated non-negative cost  $c_i$ . Design a dynamic programming solution for finding an acceptable subset  $P$  with the smallest total cost  $\sum_{i \in P} c_i$ . For full credit, your solution must run in  $O(n^2)$  time and space.

[Hint: Consider the set of all the “breakpoints”:  $\{a, b, s_1, f_1, s_2, f_2, \dots, s_n, f_n\}$ . Suppose the *distinct* breakpoints in the ascending order are  $a = p_1 < p_2 < \dots < p_m = b$  for some  $m$ . It may be useful to think of a subproblem where you want to cover the sub-interval  $[p_1, p_j]$  using only some of the guards. Do not forget to bound the maximum number of distinct breakpoints  $m$  in terms of  $n$ .]

(a) [5 Points] Define an array storing the necessary information from subproblems. Clearly define what each entry means and how you would compute the desired solution given this array.

(b) [5 Points] Write a Bellman equation and briefly justify its correctness.

(c) [2.5 Points] In what order would you compute the entries in a bottom-up implementation?

(d) [2.5 Points] Analyze the worst-case running time and space complexity of your algorithm.

### Solution to Q3

(a) Sort the guards such that  $f_1 \leq \dots \leq f_n$ . Further, as the hint suggests, sort the distinct breakpoints such that  $a = p_1 < \dots < p_m = b$ . Now, for  $0 \leq i \leq n$  and  $1 \leq j \leq m$ , define  $OPT[i, j]$  to be the smallest cost needed to cover  $[a, p_j]$  (with  $j = 1$ , i.e.,  $[a, a]$  considered trivially covered) using only the first  $i$  guards in the sorted order.

To reconstruct the optimal solution, we look at the Bellman equation below and define  $S[i, j]$  to be  $Y$  if guard  $i$  is used,  $N$  if guard  $i$  is not used, and  $\perp$  in the first two edge cases.

Then, to construct the final solution, we start  $P = \emptyset, i = n, j = m$ . Then, until  $S[i, j] = \perp$ , we do the following:

- If  $S[i, j] = Y$ , then  $P \leftarrow P \cup \{i\}, i \leftarrow i - 1, j \leftarrow k$  (where  $p_k = s_i$ ).
- If  $S[i, j] = N$ , then  $i \leftarrow i - 1$ .

At the end, we return  $P$ .

(b) The Bellman equation is as follows.

$$(OPT[i, j], S[i, j]) = \begin{cases} (0, \perp) & \text{if } j = 1, \\ (\infty, \perp) & \text{if } j \geq 2, i = 0, \\ (OPT[i - 1, j], N) & \text{if } j \geq 2, i \geq 1, p_j \notin [s_i, f_i], \\ (OPT[i - 1, j], N) & \text{if } j \geq 2, i \geq 1, p_j \in [s_i, f_i], OPT[i - 1, j] < c_i + OPT[i - 1, k], \text{ where } p_k = s_i, \\ (c_i + OPT[i - 1, k], Y) & \text{if } j \geq 2, i \geq 1, p_j \in [s_i, f_i], OPT[i - 1, j] \geq c_i + OPT[i - 1, k], \text{ where } p_k = s_i. \end{cases}$$

Note that choosing guard  $i$  can only be helpful if  $p_j \in [s_i, f_i]$ : if  $p_j < s_i$ , then the guard doesn't cover any useful portion, and if  $p_j > f_i$ , then due to the sorted order, none of guards  $1, \dots, i$  can cover point  $p_j$  (so our recursive solution will keep calling  $OPT$  with one smaller  $i$  until it reaches  $i = 0$  and returns  $\infty$ ). If choosing guard  $i$  can be helpful, then we want to consider both choosing guard  $i$  (in which case we only have interval  $[a, s_i]$  left to be covered) and not choosing guard  $i$  (in which case we still need to cover  $[a, p_j]$  with only guards  $1, \dots, i - 1$ ).

(c) Since  $(OPT[i, j], S[i, j])$  only depends on  $OPT[i - 1, \cdot]$ , we compute them in the following order: loop over  $i = 0, \dots, n$ , and for each  $i$ , loop over  $j = 1, \dots, m$ .

(d) Since there are at most  $2n$  breakpoints, we have  $m = O(n)$ . Hence, both arrays require  $O(n^2)$  space. Further, computing each array entry requires  $O(1)$  times given previous entries. Hence, the worst-case running time is  $O(n^2)$  as well.

#### Q4 [15 Points] Divide the Workload

You are the CEO of a company which employs  $n$  workers to perform  $m$  tasks. Each worker  $i$  is supposed to work a total of  $w_i$  hours and each task  $j$  requires a total of  $t_j$  hours of work. Assume that  $\sum_{i=1}^n w_i = \sum_{j=1}^m t_j$ . The floor supervisor has come up with an ideal work schedule represented as matrix  $A$ , where row  $i$  represents worker  $i$ , column  $j$  represents task  $j$ , and  $A_{i,j}$  is the number of hours worker  $i$  will spend on task  $j$ . Matrix  $A$  has the property that the sum along each row  $i$  is exactly  $w_i$  and the sum along each column  $j$  is exactly  $t_j$ .

There is just one problem. The floor supervisor has taken the liberty of using fractional values for  $A_{i,j}$ -s, forgetting the recent company policy that a worker must spend an *integral* number of hours on a task. Luckily, all the  $w_i$ -s and  $t_j$ -s are integral. Your goal is to prove that it is always possible to “round” matrix  $A$  into some matrix  $B$  while preserving the row and column sums (i.e. set each  $B_{i,j}$  to be either  $\lfloor A_{i,j} \rfloor$  or  $\lceil A_{i,j} \rceil$  such that each row  $i$  of  $B$  still sums to  $w_i$  and each column  $j$  of  $B$  still sums to  $t_j$ ). The example below shows such a rounding of a  $3 \times 3$  matrix.

$$A = \left[ \begin{array}{ccc|c} 2.6 & 0 & 0.4 & 3 \\ 0.8 & 2.9 & 1.3 & 5 \\ 1.6 & 0.1 & 5.3 & 7 \\ \hline 5 & 3 & 7 & \end{array} \right] \longrightarrow B = \left[ \begin{array}{ccc|c} 2 & 0 & 1 & 3 \\ 1 & 3 & 1 & 5 \\ 2 & 0 & 5 & 7 \\ \hline 5 & 3 & 7 & \end{array} \right]$$

(a) [2.5 Points] Consider the matrix  $A'$  obtained by replacing each entry of  $A$  with its fractional part (e.g. replacing 1.3 with 0.3, 2.6 with 0.6, 0.1 with 0.1, etc). First, argue that  $A'$  must also have *integral* row and column sums. Next, argue that if  $A'$  can be rounded while preserving the row and column sums, then  $A$  can be as well.

(b) [10 Points] Note that each  $A'_{i,j} \in [0, 1]$ ; hence, rounding it means setting it to either 0 or 1 (except, if  $A'_{i,j} \in \{0, 1\}$  then the rounding must not change its value). Using network flow techniques, show that  $A'$  can be rounded while preserving row and column sums. Justify your answer.

[Hint: Construct a network with integral edge capacities, use  $A'$  to construct a max flow with fractional flow values on edges, and then use the integrality property of the Ford-Fulkerson algorithm (i.e. that it finds a max flow in which each edge carries an integral amount of flow).]

(c) [2.5 Points] What is the worst-case running time of the naïve Ford-Fulkerson on your network?

#### Solution to Q4

(a) Let  $F$  be the matrix where  $F_{i,j} = \lfloor A_{i,j} \rfloor$ . Then,  $A' = A - F$ . Since row/column sums of both  $A$  and  $F$  are integral, the row/column sums of  $A'$  are also integral.

If  $A'$  can be rounded into  $B'$  while preserving the row/column sums, then note that  $B = F + B'$  gives a rounding of  $A$ : it has the same row/column sums as that of  $F + A' = A$  and each of its entries is either  $\lfloor A_{i,j} \rfloor + 0$  or  $\lfloor A_{i,j} \rfloor + 1$  (except it must be equal to  $\lfloor A_{i,j} \rfloor = A_{i,j}$  if  $A_{i,j}$  is an integer, i.e., if  $A'_{i,j} = 0$ ).

(b) Construct a network as follows.

- Add a source node  $s$  and a target node  $t$ .
- Add a vertex  $r_i$  for each row  $i$  and a vertex  $c_j$  for each column  $j$ .
- Add an edge  $s \rightarrow r_i$  for each  $i$  with capacity equal to the sum of row  $i$  in  $A'$ .
- Add an edge  $c_j \rightarrow t$  for each  $j$  with capacity equal to the sum of column  $j$  in  $A'$ .
- Add a unit-capacity edge  $r_i \rightarrow c_j$  for each  $(i, j)$  with  $A'_{i,j} > 0$  (i.e. those entries which can be rounded to 1).

Consider the following flow  $f$ . Every  $s \rightarrow r_i$  and  $c_j \rightarrow t$  edge is saturated, and  $f_{r_i \rightarrow c_j} = A'_{i,j}$  for each  $(i, j)$ . Note that because the capacities of  $s \rightarrow r_i$  and  $c_j \rightarrow t$  edges are the sums of entries of row  $i$  and column  $j$  in  $A'$ , flow conservation constraints are satisfied. Edge capacity constraints are also trivially satisfied. Hence,  $f$  is a valid flow. Further, since all edges leaving  $s$  are saturated, it must be a max flow. Hence, max flow value is  $\sum_{i,j} A'_{i,j}$ .

However, since the network has edges of integral capacity, the Ford-Fulkerson algorithm must return a flow  $f^*$  with integral flow values on edges. Define  $B'$  such that  $B'_{i,j} = 1$  if  $f^*_{r_i \rightarrow c_j} = 1$  and  $B'_{i,j} = 0$  otherwise. Then, due to the construction of the network,  $B'$  must be a rounding of  $A'$ .

(c) The network in part (b) has at most  $n+m+n \cdot m = O(n \cdot m)$  edges,  $n+m+2 = O(n+m)$  nodes, and max flow value of  $\sum_{i,j} A'_{i,j} \leq n \cdot m$ . Hence, the worst-case running time of the Ford-Fulkerson algorithm is  $O(n^2 m^2)$ .

### Q5 [15 Points] Linear Programming

(a) [5 Points] Convert the following linear program to the standard form. You only need to write the final answer; no justification is needed.

$$\begin{array}{ll} \max & 3x + 5y + 2z \\ \text{s.t.} & 5y + 10z \leq 3 - 2x \\ & 2x \leq 2 - 3y - z \\ & x, y \geq 0, z \in \mathbb{R} \end{array}$$

(b) [5 Points] Write the dual of the linear program from part (a). You do *not* need to write this in the standard form and no justification is needed.

(c) [5 Points] Consider the optimization problem from part (a), but change the objective function to maximizing  $f(x, y, z)$ , where

$$f(x, y, z) = \begin{cases} 3x + 5y, & \text{if } z \geq 0, \\ 3x + 5y + 2z, & \text{if } z < 0. \end{cases}$$

Note that  $f(x, y, z)$  is *not* linear, and hence, the new optimization problem is not linear as well. Nonetheless, show that it can be converted into an equivalent *linear* program. Provide this equivalent linear program in its standard form and justify the equivalence.

### Solution to Q5

(a)

$$\begin{array}{ll} \max & 3x + 5y + 2z' - 2z'' \\ \text{s.t.} & 2x + 5y + 10z' - 10z'' \leq 3 \\ & 2x + 3y + z' - z'' \leq 2 \\ & x, y, z', z'' \geq 0 \end{array}$$

(b) Both the dual of the original LP and the dual of the LP in the standard form would be acceptable in this part.

Dual of the original LP:

$$\begin{array}{ll} \min & 3a + 2b \\ \text{s.t.} & 2a + 2b \geq 3 \\ & 5a + 3b \geq 5 \\ & 10a + b = 2 \\ & a, b \geq 0 \end{array}$$

In the dual of the standard form LP, the  $10a+b = 2$  constraint would be replaced by two constraints:  $10a + b \geq 2$  and  $-10a - b \geq -2$ .

(c) We can use the trick from part (a) where we replace the unrestricted  $z$  by  $z' - z''$  with non-negative variables  $z'$  and  $z''$ , and then optimize the linear objective function  $3x + 5y - 2z''$ .

$$\begin{array}{ll} \max & 3x + 5y - 2z'' \\ \text{s.t.} & 2x + 5y + 10z' - 10z'' \leq 3 \\ & 2x + 3y + z' - z'' \leq 2 \\ & x, y, z', z'' \geq 0 \end{array}$$

The idea is that if the optimal solution of the original program has  $z \geq 0$ , then the corresponding optimal solution in the new LP will set  $z'' = 0$ , making the objective  $3x + 5y$ . And if the optimal solution of the original program has  $z < 0$ , then the corresponding optimal solution in the new LP will set  $z'' = -z$ , making the objective  $3x + 5y - 2z'' = 3x + 5y + 2z$ .

Formally, we can show that the optimal values of the two programs are equal by showing that each is at least the other. If  $(x, y, z)$  is an optimal solution of the original program, then note that  $(x, y, z', z'')$  is a feasible solution of the new LP with the same objective value, where, if  $z \geq 0$  then  $z' = z$  and  $z'' = 0$ , and if  $z < 0$ , then  $z' = 0$  and  $z'' = -z$ . Similarly, if  $(x, y, z', z'')$  is an optimal solution of the new LP, then  $(x, y, z = z' - z'')$  is a feasible solution of the original program. To claim that it has the same objective value, we need to show that  $z'' > 0$  implies  $z' = 0$ . This is true because if both are positive, then reducing both by a small amount  $\delta$  yields a feasible solution with a better objective value, a contradiction.

### Q6 [20 Points] SAT

Recall that a CNF formula  $\varphi = C_1 \wedge \dots \wedge C_m$  is a conjunction of clauses, where each clause is a disjunction of (any number of) literals. Recall the NP-complete problem SAT.

**SAT:**

**Input:** A CNF formula  $\varphi$ .

**Question:** Does  $\varphi$  have a satisfying assignment?

Now, consider the following two variants of it.

**TripleSAT:**

**Input:** A CNF formula  $\varphi$ .

**Question:** Does  $\varphi$  have at least *three* different satisfying assignments?

**TwoThirdsSAT:**

**Input:** A CNF formula  $\varphi$ .

**Question:** Is there an assignment satisfying at least two-thirds ( $2/3$ ) of the clauses of  $\varphi$ ?

(a) [3 Points] Prove that TripleSAT is in NP.

(b) [7 Points] Prove that TripleSAT is NP-hard through a reduction from SAT.

(c) [3 Points] Prove that TwoThirdsSAT is in NP.



(d) [7 Points] Prove that TwoThirdsSAT is NP-hard through a reduction from SAT.

### Solution to Q6

(a) We can provide, as advice, three different satisfying assignments of  $\varphi$ .

(b) Given an instance  $\varphi$  of SAT, construct an instance  $\varphi'$  of TripleSAT by adding two fresh variables  $x_1, x_2$  and adding a clause  $(x_1 \vee x_2)$ . Note that satisfying assignments of  $\varphi'$  are formed by taking satisfying assignments of  $\varphi$  and appending  $(x_1, x_2) = (T, T), (T, F),$  or  $(F, T)$ . Hence, the number of satisfying assignments of  $\varphi'$  is exactly three times the number of satisfying assignments of  $\varphi$ , which implies that  $\varphi'$  has at least three satisfying assignments if and only if  $\varphi$  has a satisfying assignment, implying that both instances have the same answer.

(c) We can provide, as advice, an assignment that satisfies at least two-thirds of the clauses of  $\varphi$ .

(d) Given an instance  $\varphi$  of SAT with  $n$  variables and  $m$  clauses, construct an instance  $\varphi'$  of TwoThirdsSAT by adding  $m$  fresh variables  $x_1, \dots, x_m$  and  $2m$  clauses  $x_1, \bar{x}_1, \dots, x_m, \bar{x}_m$ . Note that  $\varphi'$  has  $3m$  clauses and every assignment satisfies exactly  $m$  of the  $2m$  newly added clauses. Hence, an assignment of  $\varphi'$  satisfies at least  $2m$  of its clauses if and only if the corresponding assignment of  $\varphi$  satisfies all its  $m$  clauses. Hence, both instances have the same answer.

### Q7 [15 Points] Sabotage!

There is an undirected graph  $G = (V, E)$ , where nodes in  $V$  are servers and edges in  $E$  are cables running between pairs of servers. A set of  $k > 2$  servers  $S = \{v_1, \dots, v_k\} \subseteq V$  is trying to collaboratively solve a problem and you want to sabotage this!

Specifically, you want to remove a subset of edges  $T \subseteq E$  such that all nodes in  $S$  become disconnected from one another (i.e. there is no path left between any two of them). You want  $|T|$  to be as small as possible. Consider the following greedy algorithm.

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#### Algorithm 3: Greedy-Sabotage

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```

1 for  $i = 1, \dots, k$  do
2   Let  $E_i$  be the smallest subset of edges we need to remove to disconnect  $v_i$  from every
   other node in  $S$ . (It turns out that  $E_i$  can be computed efficiently, but do not worry
   about this.)
3 end
4 Remove the union of  $k - 1$  smallest  $E_i$ -s (i.e. the union of all but the largest  $E_i$ ).
```

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(a) [5 Points] Prove that Greedy-Sabotage returns a feasible solution  $T$  (i.e. removing the set of edges  $T$  it returns will indeed disconnect every pair of nodes in  $S$ ).

(b) [10 Points] Prove that Greedy-Sabotage achieves a  $2 - 1/k$  approximation ratio. For partial credit, prove a slightly weaker approximation ratio of 2.

[Hint: Let  $T^*$  denote the optimal solution. For each  $i \in \{1, \dots, k\}$ , let  $V_i \subseteq V$  denote the set of nodes in the connected component containing  $v_i$  (but no other node in  $S$ ) that remains after removing  $T^*$ . Can you relate the number of edges in  $T^*$  with one endpoint in  $V_i$  with  $|E_i|$ ?

### Solution to Q7

(a) Let  $T$  be the set of edges removed by Greedy-Sabotage. Consider any two nodes  $a, b \in S$ . Since  $T$  is the union of  $k-1$  of the  $E_i$ -s, we have that either  $E_a \subseteq T$  or  $E_b \subseteq T$ . Hence, either  $a$  or  $b$  must be disconnected from all other nodes in  $S$  after removal of  $T$ , which implies that  $a$  and  $b$  are not connected to each other after removal of  $T$ . Hence, Greedy-Sabotage returns a feasible solution.

(b) Let  $E_i^*$  be the number of edges in  $T^*$  with one endpoint in  $V_i$ . Since removing  $E_i^*$  disconnects  $V_i$  from the rest of the graph, it also disconnects  $v_i$  from the other nodes in  $S$ . Since  $E_i$  is the smallest set that does this, we have  $|E_i^*| \geq |E_i|$ .

Note that  $2T^* \geq \sum_i |E_i^*| \geq \sum_i |E_i|$ , where the first transition holds because each edge in  $T^*$  is counted at most twice (once for each endpoint) in the sum on the RHS. Since  $T$  omits the  $E_i$  with the largest cardinality (which must have cardinality at least  $(1/k) \sum_i |E_i|$ ), we have that  $T \leq (1 - 1/k) \cdot \sum_i |E_i|$ . Hence,  $2(1 - 1/k) \cdot T^* \geq T$ , which means our greedy algorithm achieves an approximation factor of  $2 - 2/k$  (which is actually better than  $2 - 1/k$ ).