# CSC373 Homework 2

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# Question 1 Investment Strategy (CLRS Exercise 15-10)

(a) Suppose for contradiction that there does not exist an optimal strategy that in each year, puts all money into a single investment.

Consider an optimal strategy S\*. Without loss of generation, assume it invest  $d_1$  dollars in investment  $i_i$ , and invest  $d_2$  dollars in investment  $i_2$  in year  $j_1$ , and from  $j_1$  to  $j_n$  years the money hasn't been switched. Assume compound interests.

Then the profit for  $i_1$  is  $d_1(\prod_{j=j_1}^{j_n}(1+r_{i_1j})-1)$ , and the profit for  $i_2$  is  $d_2(\prod_{j=j_1}^{j_n}(1+r_{i_2j})-1)$ . Without loss of generation, assume  $(r_{i_1j}\prod_{j=j_1}^{j_n}(1+r_{i_1j})-1) \geq (\prod_{j=j_1}^{j_n}(1+r_{i_2j})-1)$ . Then we can invest all  $d_1+d_2$  dollars into investment  $i_i$  to win more or equivalent profit, while reduces the number of investment from year  $j_1$  to  $j_n$  by 1.

Therefore, by repetitively doing this way, we can reduce optimal strategy  $S^*$  to only involve one investment each year. This is a contradiction. Thus, there exists an optimal strategy that in each year, puts all money into a single investment.

(b) Consider an optimal solution OPT. In year j that j < n, there are two options regarding how to invest in the following year j + 1:

Option 1. Do not move the money, pay a fee of  $f_1$  dollars.

Option 2. Shift money to other investments by either shifting money between existing investments or to a new investment, pay a fee of  $f_2$  dollars.

OPT is the best of the two options that gives the most return from year j + 1 to n.

Therefore the problem of planning optimal investment strategy exhibits optimal substructure.

(c) Let  $OPT(j,i) = \max$  return from year 1 to j, and choose investment i at year j.

Base Case: at year 1, choose i that maximizes  $r_{i1}$ , and thus the max return is  $r_{i1}$ .

Two cases regarding year j:

1.optimal return is that money is not moved in year j, which means this year's investment will be the same as previous year which is  $i_{j-1}$ .

2.optimal return is that money is moved in year j. Choose this year's investment from  $i = 1, ..., n, i \neq i_{j-1}$  that gives the maximum return for year j.

Bellman equation:

```
OPT(j,i) = \max\{OPT(j-1,i_{j-1})(1+r_{i_{j-1}j}) - f_1, OPT(j-1,i_{j-1})(1+\max_{i\neq i_{j-1}}(r_{ij})) - f_2\}
```

### Algorithm 1 Store maximum return at end of each year and the investment at each year to OPT

```
1: procedure StoreInvest(d, n, i_1, ..., i_n, r_{11}, ..., r_{n10}, f_1, f_2)
         Initialize OPT as a list of zero tuples [(0,0),(0,0),...(0,0)]
 3:
         OPT[1] \leftarrow (d(1 + max(r_{i1})), max\_index(r_{i1}))
         for j = 2, ..., 10 do
 4:
             i_{max} \leftarrow max\_index(r_{ij})
 5:
             i_{j-1} \leftarrow OPT[j-1][1]
 6:
             returns = [OPT[j-1][0](1+r_{i_{j-1}j}) - f_1, OPT[j-1][0](1+r_{i_{max}j}) - f_2]
 7:
             invests = [i_{max}, i_{j-1}]
 8:
             if i_{max} \geq i_{j-1} then
 9:
                  idx = max\_index(returns)
10:
             else
11:
12:
                  idx = 1
         \begin{array}{l} OPT[j][0] \leftarrow returns[idx], OPT[j][1] \leftarrow invests[idx] \\ \textbf{return} \ OPT \end{array}
13:
```

Then we can find the optimal investment at each year j from OPT[j][1], and the maximum return at end of 10 years from OPT[10][0].

In the function StoreInvest, There are 10 iterations in the for loop in total. At each iteration, line 5 cost O(n), line 6 to 10 cost O(1). To find the optimal investment and maximum return cost O(1). Therefore the runtime in total is O(n).

(d) With the additional restriction, there is no longer guaranteed that regarding each year j, there would be exactly two cases. Because if we exceed the 15,000 in the case which gives the maximum return, we need to reconsider a best possible investment. If the reconsidered best choice also exceed the limit, we need to repeatedly do the step again.

In addition, whether to switch money at the beginning of each year is now dependent on how much money we have at the beginning of each year. If we exceed the limit, then we must switch money.

Therefore, there is no clear bound that how many sub-problems we have in each year. Thus the optimal substructure doesn't exist.

## Question 2 Party Planning

- (a) Let  $OPT(T) = \text{maximum sum of "fun" ratings of the guests from the tree T, based on the invitation rule in the problem description.$ 
  - OPT(T) is the quantity that dynamic programming approach computes.
- (b) Bellman:

```
OPT(T) = \{ \begin{array}{c} rating(root) & \text{T's root has no children} \\ \max\{\sum_{t \in child(T)} OPT(t), rating(root) + \sum_{t \in grandchild(T)} OPT(t)\} \end{array} \right.
```

Justification:

Base Case: when T only has one node, then the maximum sum of "fun" rating can only be rating(root). Two cases regarding when T has subtrees:

- 1. We do not invite the root of T, then we find the optimal sum of "fun" ratings from those children trees.
- 2. We invite the root of T, then the direct children of its root cannot be invited, from the invitation rule in the problem description. We can only invite from its children of children, and find the optimal sum of "fun" ratings from those grandchildren trees.
- (c) The Bellman equation in part (b) is computed in Top-Down order.

Let r is the root the whole tree of the company structure. The R(r) is initialized beforehand as rating(r), since from the invitation rule, the president must attend the party, and rating(r) is a negative number. Other R(v) are initialized as 0.

Call ComputeRatings(r), then the algorithm will calculate and store the maximum ratings of tree rooted at each node v in R(v), in the Top-Down order.

#### **Algorithm 2** Store maximum ratings of tree rooted at node v in R(v)

```
1: procedure ComputeRatings(v)
       if R(v) was not initialized then
          if child(v) is None then
3:
4:
              R(v) = rating(v) return R(v)
5:
          sum1 \leftarrow 0
          for c in child(v) do
6:
              if R(c) was initialized then
7:
                 sum1 = sum1 + R(c)
8:
              else
9:
                 sum1 = sum1 + ComputeRatings(c)
10:
          sum2 \leftarrow rating(v)
11:
          for c in child(child(v)) do
12:
13:
              if R(c) was initialized then
                 sum2 = sum2 + R(c)
14:
              else
15:
                 sum2 = sum2 + ComputeRatings(c)
16:
          R(v) \leftarrow max(sum1, sum2)
17:
        return R(v)
```

(d) Assume the total number of people in the company is n.

time complexity:

In the first function call ComputeRatings(r), in the first for-loop from line 6 to 10, there is the first set of possible recursive calls, and the maximum number of possible recursive calls in this for-loop is O(n). Similarly, in the second for-loop from line 12 to 16, the maximum number of possible recursive calls in this for-loop is O(n).

Since we omit the unnecessary repeated recursive calls by storing R(v) and checking their initialization, the maximum rating at each node is calculated exactly once.

Therefore, there will be maximum n function calls in total, to calculate and store all R(v).

Thus, the total runtime is O(n).

space complexity:

If we assume we need 1 bit to store each element in the list R(v), then the algorithm need maximum O(n)

bits to store all information of maximum ratings at each node, in order to calculate the maximum ratings of the whole tree.

(e) Initialize L as a global list.

Call ProduceList(r,R) will produce the guest list , where R is the list of maximum ratings calculated before, and r is the root of the whole tree.

#### **Algorithm 3** Produce the guest list L from the maximum ratings given by R

```
1: procedure ProduceList(v, R)
        if v is the root of whole tree or child(v) is None then
             L.append(v) return
 3:
        sum1 \leftarrow 0
 4:
        for c in child(v) do
 5:
             sum1 = sum1 + R(c)
 6:
        sum2 \leftarrow rating(v)
 7:
        \mathbf{for}\ c\ \mathrm{in}\ \mathrm{child}(\mathrm{child}(v))\ \mathbf{do}
 8:
             sum2 = sum2 + R(c)
 9:
        if R(v) = sum1 then
10:
             \mathbf{for} \ c \ \mathrm{in} \ \mathrm{child}(v) \ \mathbf{do}
11:
                 ProduceList(c, R)
12:
        if R(v) = sum2 then
13:
14:
             L.append(v)
             for c in child(child(v)) do
15:
                 ProduceList(c, R)
16:
        return
```

# Question 3 Doctor Scheduling

- (a) Let the days be  $d_1, d_2, \dots, d_n$ . Each doctor can work at most n days. Create a flow network instance with the set of vertices  $\{s, t, a_1, \dots, a_n, d_1, \dots, d_n\}$  and the following edges:
  - $\{s, a_i\}$  with capacity n, for each  $a_i$
  - $\{d_j, t\}$  with capacity  $a_j$ , for each  $d_j$
  - $\{a_i, d_j\}$  with unit capacity, for each  $a_i$  and  $d_j$  such that doctor  $a_i$  would like to work on day  $d_j$  (i.e.  $d_j \in A_i$ ).

The algorithm is as follows:

- (i) Compute a maximum flow f in the above instance via the Ford-Fulkerson algorithm, which will be an integral flow.
- (ii) If  $0 \le f(s, a_i) \le n$  for each doctor  $a_i$ , and  $f(d_j, t) = a_j$  for each day  $d_j$ , then a feasible schedule exists. Assign doctor  $a_i$  to work on day  $d_j$  whenever  $f(a_i, d_j) = 1$ .
- (iii) Otherwise, report that no schedule is possible given the doctors' availability.
- (b) To prove that the algorithm is correct, we will establish a 1-1 correspondence between our integral flows f and feasible schedules, and the correspondence will be that doctor  $a_i$  work on day  $d_j \Leftrightarrow f(a_i, d_j) = 1$ . We will prove this in both directions.
  - Valid integral flow  $\Rightarrow$  feasible schedule: Take any feasible integral flow f and construct the corresponding schedule with the above correspondence. We want to prove that this is a feasible schedule. Each  $d_j$  must have an outgoing flow of  $a_j$ . Since the edges from doctors to days have unit capacity and the flow is integral, each  $d_j$  has  $a_j$  outgoing edges carrying a unit flow, and the rest carrying no flow. Since  $a_j \leq n$  and each edge  $\{a_i, d_j\}$  satisfies  $d_j \in A_i$ , each  $a_i$  has at most  $|A_i|$  outgoing edges, and  $|A_i| \leq n$ . Therefore, by our correspondence, each doctor  $a_i$  only work when they are available and exactly  $a_j$  doctors work on day  $d_j$ , so the schedule is feasible.
  - Feasible schedule  $\Rightarrow$  valid integral flow: Take any feasible schedule and construct a integral flow f with the above correspondence. Since there are  $a_j$  doctors working on each day  $d_j$ , each  $d_j$  has an outgoing flow of  $a_j$ . And each doctor  $a_i$  only work on days when they are available, so each  $a_i$  has an incoming flow  $l_i \leq |A_i| \leq n$ . Therefore,  $f(s, a_i) = n$  for each  $a_i$  and  $f(d_j, t) = a_j$  for each  $d_j$  satisfies the flow conservation constraints, and the flow is valid.

### Runtime:

This flow network has n+n+2 vertices and  $n+n+\sum_{i=1}^{n}|A_i|\leq 2n+n^2$  edges.

The sum of capacities of edges leaving s is  $\sum_{i=1}^{n} l_i \leq n^2$ , since  $l_i \leq |A_i| \leq n$  for each i. Hence, the worst-case running time of the algorithm is  $\mathcal{O}(n^4)$ .

(c) Assume  $c \le n - \min_{1 \le i \le n} |A_i|$ .

Create a flow network instance with the set of vertices  $\{s, t, a_1, \dots, a_n, a'_1, \dots, a'_n, d_1, \dots, d_n\}$  and the following edges:

- $\{s, a_i\}$  with capacity n, for each  $a_i$ , and  $\{s, a_i'\}$  with capacity c, for each  $a_i'$
- $\{d_j, t\}$  with capacity  $a_j$ , for each  $d_j$
- $\{a_i, d_j\}$  with unit capacity, for each  $a_i$  and  $d_j$  such that doctor  $a_i$  would like to work on day  $d_j$  (i.e.  $d_j \in A_i$ ).
- $\{a'_i, d_i\}$  with unit capacity, for each  $a_i$  and  $d_j$  such that  $d_j \notin A_i$ .

The algorithm is as follows:

- (i) Compute a maximum flow f in the above instance via the Ford-Fulkerson algorithm, which will be an integral flow.
- (ii) If  $0 \le f(s, a_i) \le n$  and  $0 \le f(s, a_i') \le c$  for each doctor  $a_i$ , and  $f(d_j, t) = a_j$  for each day  $d_j$ , then a feasible schedule exists. Assign doctor  $a_i$  to work on day  $d_j$  whenever  $f(a_i, d_j) = 1$  or  $f(a_i', d_j) = 1$ .
- (iii) Otherwise, report that no schedule is possible given the doctors' availability.
- (d) To prove that the algorithm is correct, we will establish a 1-1 correspondence between our integral flows f and feasible schedules. We will prove this in both directions.

• Valid integral flow  $\Rightarrow$  feasible schedule: Take any feasible integral flow f and construct the corresponding schedule with the above correspondence. We want to prove that this is a feasible schedule. Each  $d_j$  must have an outgoing flow of  $a_j$ . Since the edges from doctors to days have unit capacity and the flow is integral, each  $d_j$  has  $a_j$  outgoing edges carrying a unit flow, and the rest carrying no flow. Since each edge  $\{a_i, d_j\}$  satisfies  $d_j \in A_i$ , each  $a_i$  has at most  $|A_i|$  outgoing edges; and each edge  $\{a'_i, d_j\}$  satisfies  $d_j \notin A_i$ , each  $a'_i$  has at most  $(n - |A_i|)$  outgoing edges, and  $n - |A_i| \le c$ .

Therefore, by our correspondence, each doctor  $a_i$  only work when they are available or at most c days outside their availability list, and exactly  $a_j$  doctors work on day  $d_j$ , so the schedule is feasible.

• Feasible schedule  $\Rightarrow$  valid integral flow: Take any feasible schedule and construct a integral flow f with the above correspondence. Since there are  $a_j$  doctors working on each day  $d_j$ , each  $d_j$  has an outgoing flow of  $a_j$ . And each doctor  $a_i$  work on days when they are available or at most c days outside their availability list, so each  $a_i$  has an incoming flow  $l_i \leq |A_i| \leq n$ , and each  $a_i'$  has an incoming flow  $l_i' \leq c \leq n - |A_i|$ . Therefore,  $f(s, a_i) = n$  for each  $a_i$ ,  $f(s, a_i') = c$  for each  $a_i'$ , and  $f(d_i, t) = a_i$  for each  $d_i$  satisfies the flow conservation constraints, and the flow is valid.

#### Runtime:

This flow network has 2n + n + 2 vertices and  $2n + n + \sum_{i=1}^{n} |A_i| + \sum_{i=1}^{n} |A - A_i| \le 3n + n^2$  edges. The sum of capacities of edges leaving s is  $\sum_{i=1}^{n} l_i + l'_i \le \sum_{i=1}^{n} |A_i| + n - |A_i| \le n^2$ . Hence, the worst-case running time of the algorithm is  $\mathcal{O}(n^4)$ .

### Question 4 Exam Invigilation

- (a) Let the CPOs be  $CPO_1, CPO_2, \cdots, CPO_n$ , and the exam slots be  $e_1, e_2, \cdots, e_m$ , where each exam slot  $e_i$  has  $l_i$  exams. Assume each  $CPO_i$  has a list of availability  $A_i$ , i.e.  $CPO_i$  can invigilate an exam during exam slot  $e_j$  if  $e_j \in A_i$ , and the maximum number of exam slots that  $CPO_i$  can invigilate is  $c_i, c_i \leq |A_i|$ . Since there are m exams, there are at most m days. Let  $d_{i1}, d_{i2}, \cdots, d_{im}$  be the days of exams for  $CPO_i$ . Create a flow network instance with the set of vertices  $\{s, t, CPO_1, \cdots, CPO_n, d_{11}, \cdots, d_{1m}, \cdots, d_{n1}, \cdots, d_{nm}, e_1, \cdots, e_m\}$  and the following edges:
  - $\{s, CPO_i\}$  with capacity  $c_i$ , for each  $CPO_i$
  - $\{CPO_i, d_{ij}\}$  with capacity 2, for each  $CPO_i$  and  $d_{ij}$  such that  $CPO_i$  has availability on day j.
  - $\{d_{ij}, e_k\}$  with capacity 1, for each  $d_{ij}$  and  $e_k$  such that exam  $e_k$  is on day j and  $CPO_i$  is available to invigilate  $e_k$ .
  - $\{e_k, t\}$  with capacity  $[1.1l_k]$ , for each exam slot  $e_k$

The algorithm is as follows:

- (i) Compute a maximum flow f in the above instance via the Ford-Fulkerson algorithm, which will be an integral flow.
- (ii) If  $0 \le f(s, CPO_i) \le c_i$  for each  $CPO_i$ , and  $f(e_k, t) = \lceil 1.1l_k \rceil$  for each exam slot  $e_k$ , then a feasible schedule exists. Assign  $CPO_i$  to invigilate on exam period  $e_k$  whenever  $f(d_{ij}, e_k) = 1$ .
- (iii) Otherwise, report that no assignment is possible given the CPOs' availability.
- (b) To prove that the algorithm is correct, we will establish a 1-1 correspondence between our integral flows f and feasible assignments. We will prove this in both directions.
  - Valid integral flow  $\Rightarrow$  feasible assignment: Take any feasible integral flow f and construct the corresponding assignment with the above correspondence. We want to prove that this is a feasible assignment. Each  $e_k$  must have an outgoing flow of  $\lceil 1.1l_k \rceil$ . Since the edges from day per CPO  $d_{ij}$  to exam periods have unit capacity and the flow is integral, each  $e_k$  has exactly  $\lceil 1.1l_k \rceil$  incoming edges carrying a unit flow. And since the flow satisfies  $0 \le f(s, CPO_i) \le c_i$ , each CPO will not invigilate more exams than their maximum availability. Since  $f(d_{ij}, e_k) \le 2$ , each CPO will invigilate at most two exams on a single day. Therefore, by our correspondence, each  $CPO_i$  only invigilate the exam periods when they are available, so the assignment is feasible.
  - Feasible assignment  $\Rightarrow$  valid integral flow: Take any feasible assignment and construct a integral flow f with the above correspondence. By the construction, it is clear that f is integral. Each  $CPO_i$  only invigilate on exam slots when they are available and the number will not exceed the maximum available number of exam slots, so  $f(s, CPO_i) \leq c_i$  for each  $CPO_i$ . Each CPO will invigilate at most

2 exams on each day, so  $d_{ij} \leq 2$ . And each exam slot has  $\lceil 1.1l \rceil$  available CPOs, so it has  $\lceil 1.1l \rceil$  incoming edges carrying a unit flow, so  $f(e_k,t) = \lceil 1.1l_k \rceil$  for each  $e_k$ . Therefore, the flow satisfies the flow conservation constraints, and it is valid.

#### Runtime:

This flow network has n + mn + m + 2 vertices. There are n CPOs and each has at most m edges to their corresponding days. There are mn days per CPO, and each day per CPO has at most 3 edges to exam slots. So the number of edges is at most n + mn + 3mn + m = n + m + 4mn edges.

The sum of capacities of edges leaving s is  $\sum_{i=1}^{n} c_i \leq n^2$ . Hence, the worst-case running time of the algorithm is  $\mathcal{O}(mn^3)$ .