Algorithm HW#2

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1 Problem 1.

First we define some notation. $\mathcal{N} = \{1, 2, ..., n\}$ would be the elements and $\mathcal{S} = \{S_1, ..., S_m\}$ would be the sets in the set cover problem, where each $S_i \subseteq \mathcal{N}$. Let $w_i = w(S_i)$ be the cost to choose S_i . The corresponding relaxed LP is:

minimize
$$\sum_{S_i \in \mathcal{S}} w_i x_i$$

s.t. $\sum_{j \in S_i} x_i \ge 1, \quad \forall j \in \mathcal{N}$ (1)

The dual LP is

maximize
$$\sum_{j \in \mathcal{N}} \alpha_j$$
s.t.
$$\sum_{j \in S_i} \alpha_j \le w_i, \quad \forall S_i \in \mathcal{S}$$
 (2)

The greedy algorithm mentioned in class could be extend to give a dual feasible solution.

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Algorithm 1: The greedy algorithm taught in class

1 C \leftarrow \mathcal{N} // Elements that has not been covered

2 while C \neq \emptyset do

3 | Find a set S with S \cap C \neq \emptyset and minimize t = w(S)/|S \cap C|.

4 | Set \alpha_j \leftarrow t/H_k, for each j \in S \cap C. /* j is covered in this round */

5 | Choose S. /* The cost is w(S) = t|S \cap C| = \sum_{j \in S \cap C} H_k \alpha_j */

6 | C \leftarrow C \setminus S.
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If this algorithm indeed gives a feasible dual solution $\alpha = (\alpha_i)$, then it is easy to see that the algorithm would give a solution of the set cover problem with cost equal to $H_k \sum \alpha_i$, and

$$H_k \sum \alpha_j \le H_k \text{ OPT}_{\text{LP,dual}} = H_k \text{ OPT}_{\text{LP,primal}} \le H_k \text{ OPT}_{\text{ILP,primal}}$$

Thus it is an H_k -approximation solution.

Now, for any set $S = S_i$, we shall prove that $\sum_{j \in S} \alpha_j \leq w(S)$. Assume that $S = \{e_1, e_2, \dots, e_h\}, h \leq k$, and let t_i be the time (round) which e_i is covered. Reorder the

element so we could assume that $t_1 \leq t_2 \leq \cdots \leq t_h$. Just before e_i is choosen, if we choose S to cover e_i , the cost is w(S)/(h-i+1), thus we must have $\alpha_{e_i} \leq w(S)/((h-i+1)H_k)$. Hence

$$\sum \alpha_{e_i} \le w(S) H_k \left(\frac{1}{1} + \frac{1}{2} + \dots + \frac{1}{h} \right) \le w(S) H_h / H_k \le w(S)$$

and the prove is completed.

2 Problem 2.

Similar to problem 1., the corresponding relaxed LP is now:

minimize
$$\sum_{S_i \in \mathcal{S}} w_i x_i$$
s.t.
$$\sum_{j \in S_i} x_i \ge k_j, \quad \forall j \in \mathcal{N}$$
 (3)

The dual LP is

maximize
$$\sum_{j \in \mathcal{N}} k_j \alpha_j$$
s.t.
$$\sum_{j \in S_i} \alpha_j \le w_i, \quad \forall S_i \in \mathcal{S}$$
 (4)

We make a small modification to the algorithm.

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Algorithm 2: The modified greedy algorithm
                 // Elements that has not been covered
 2 foreach j \in \mathcal{N} \ \mathbf{do} \ \alpha_j \leftarrow 0, \ y_j \leftarrow 0 // y_j is the number of time j is covered
 з while C \neq \emptyset do
       Find a set S with S \cap C \neq \emptyset and minimize t = w(S)/|S \cap C|.
       Define D \triangleq S \cap C.
                                           /* We say j is covered once in this round */
       foreach j \in D do
           \alpha_j \leftarrow \alpha_j + t/(H_n k_j)
        y_j \leftarrow y_j + 1 if y_j = k_j then /* We say j is fully covered in this round */
           end
11
       end
12
       Choose S.
13
14 end
```

In each round The cost of choosing S is

$$w(S) = w(S)|D|/|D| = H_n \sum_{j \in D} k_j t/(H_n k_j) = H_n \sum_{j \in D} k_j \Delta \alpha_j,$$

where $\Delta \alpha_j$ is the amount of change of α_j in that round. Thus the algorithm give a solution of the set cover problem with cost $H_n \sum k_j \alpha_j$.

So again, if we could prove that $\alpha = (\alpha_i)$ is a feasible solution to the dual problem (6), then

$$H_n \sum \alpha_j \leq H_n \operatorname{OPT}_{\operatorname{LP,dual}} = H_n \operatorname{OPT}_{\operatorname{LP,primal}} \leq H_n \operatorname{OPT}_{\operatorname{ILP,primal}},$$

and the solution would be an H_n -approximation solution.

Similarly, for any set $S = S_i$, Assume that $S = \{e_1, e_2, \ldots, e_h\}$, and let t_i be the time (round) which e_i is **fully covered**. Reorder the element so we could assume that $t_1 \leq t_2 \leq \cdots \leq t_h$. Notice that each element e_i would be **covered once** exactly k_{e_i} times, and these all happen before e_i is fully covered.

Consider once in this k_{e_i} times when e_i is covered once. Let $j \triangleq e_i$. The change of α_{e_i} is $\Delta \alpha_{e_i} = t/H_n k_{e_i}$ where t is defined in line 4 of algorithm 2. If we choose S, then $w(S)/|S \cap C| \leq w(S)/(h-i+1)$. Thus by the minimality of t, we must have $t \leq w(S)/(h-i+1)$. Hence

$$\alpha_{e_i} = \sum \Delta \alpha_{e_i} \le k_{e_i} \cdot \frac{w(S)}{(h-i+1)H_n k_{e_i}} = \frac{w(S)}{h-i+1} H_n^{-1}$$

So

$$\sum \alpha_{e_i} \le \sum_{i} \frac{w(S)}{h - i + 1} H_n^{-1} = w(S) \frac{H_h}{H_n} \le w(S)$$

And we conclude that α is indeed a feasible solution.

3 Problem 3.

The relaxed LP correspond to the vertex cover is:

minimize
$$\sum_{v \in V} w_v x_v$$
s.t.
$$x_u + x_v > 1, \quad \forall (u, v) \in E$$
(5)

Let E(v) be all the edges adjacent to v, the dual LP is

maximize
$$\sum_{e \in E} \alpha_e$$
s.t.
$$\sum_{e \in E(v)} \alpha_e \le w_v, \quad \forall v \in V$$
 (6)

Investigating the complementary slackness, we have

$$\sum_{e} \alpha_e \le \sum_{e} {}^{1}\alpha_e (x_u + x_v) = \sum_{v} x_v \sum_{e \in E(v)} \alpha_e \le \sum_{v} w_v x_v$$

 $^{^{1}(}u,v)=e$ here

Clearly we only need to consider the solution which every $x_v \leq 1$, so we can assume $x_v + x_u \leq 2$, $\forall (u, v) \in E$. Then if either $x_v = 0$ or $\sum_{e \in E(v)} \alpha_e = w_v$ for each $v \in V$, we would have

$$\sum_{v} w_v x_v = \sum_{v} x_v \sum_{e \in E(v)} \alpha_e = \sum_{e} \alpha_e (x_u + x_v) \le 2 \sum_{e} \alpha_e$$

Thus if these an algorithm satisfy

- 1. It gives a valid vertex cover S.
- 2. It induces a feasible dual solution $\alpha = (\alpha_e)$.
- 3. The complementary slackness relation is satisfied. That is, for each vertex v, either v is not picked into S (i.e., $x_v = 0$), or $\sum_{e \in N(v)} \alpha_e = w_v$.

Then this algorithm is a 2-approximation algorithm. So we check these 3 conditions for both algorithms.

3.1 Algorithm 1

- 1. It gives a valid vertex cover: Since eventually each edge is removed, and we only remove covered edge, all edges are covered when the algorithm ends.
- 2. It induces a feasible dual solution: Let $\alpha_e = 0$ at the beginning. Define the slackness \tilde{r}_v of a vertex v as $\tilde{r}_v \triangleq w_v \sum_{e \in N(v)} \alpha_e$, Then $\tilde{r}_v = w_v = r_v$ at the very start. Now, when reducing the residual weight r_v by $c \cdot \deg(v)$, we also increase α_e by c for all edges e. Then it is easy to see that \tilde{r}_v also decrease by $c \cdot \deg(v)$. Thus $r_v = \tilde{r}_v$ in any moment. Since $c \cdot \deg(v) = \min r_v / \deg(v) \cdot \deg(v) = \min r_v$ in each round, we never decrease r_v below zero, thus $\tilde{r}_v \geq 0$ at the end. Hence $\sum_{e \in N(v)} \alpha_e \leq w_v$ for each v when the algorithm ends, and α is then a feasible dual solution.
- 3. The complementary slackness relation is satisfied: It is clear from the statement since we add v into S only when $r_v = \tilde{r}_v = 0$ (i.e., $w_v = \sum_{e \in N(v)} \alpha_e$).

3.2 Algorithm 2

- 1. It gives a valid vertex cover: It is clear since the algorithm stops only when each edge is covered.
- 2. It induces a feasible dual solution: Similar as in algorithm 1, let $\alpha_e = 0$ at the beginning. When reducing the residual weight r_u and r_v by $t = \min(r_u, r_v)$, also increase α_e by t. Then it is easy to see that \tilde{r}_u, \tilde{r}_v also decrease by t, thus \tilde{r}_v

matches r_v . Also $t \leq r_u$ and $t \leq r_v$, so we never decrease r_v below zero. Hence $\sum_{e \in N(v)} \alpha_e \leq w_v$ and we know that α is a feasible solution.

3. The complementary slackness relation is satisfied: It is clear from the statement since we add v into S only when it has zero residual weight r_v .

Collaborators: None.

4 Problem 4.

The relaxed LP correspond to the metric uncapacitated facility locating problem with penalty is:

minimize
$$\sum c_{i,j} x_{i,j} + \sum f_i y_i + \sum p_j z_j$$
s.t.
$$z_j + \sum_i x_{i,j} \geq 1, \quad \forall j$$

$$y_i - x_{i,j} \geq 0, \quad \forall i, j$$

$$y_i, x_{i,j}, z_j \geq 0, \quad \forall i, j$$

$$(7)$$

The dual LP is:

maximize
$$\sum \alpha_{j}$$

s.t. $\alpha_{j} - \beta_{i,j} \leq c_{i,j}, \forall i, j$
 $\sum_{j} \beta_{i,j} \leq f_{i}, \forall i$
 $\alpha_{j} \leq p_{j}, \forall j$
 $\alpha_{j}, \beta_{i,j} \geq 0, \forall i, j$

$$(8)$$

Thus we modify the algorithm taught in class as following, where some definition is listed below:

- If we choose to give up client j, we say j is **covered** but **unconnected**.
- If a client j is **connected**, it is simultaneously **covered**.
- If $\beta_{i,j} > 0$, then we say client j pays for facility i.
- If there is a client j such that j **pays** for both i, i', then we say i, i' are **conflicted** with each other.
- At the end of the algorithm, we choose the facilities that is **truely opened**, and connect each **connected** client j to a facility i such that (i, j) is **truely connected**.

```
Algorithm 3: The modified greedy algorithm
   // Phase 1
1 Initially set \alpha_i \leftarrow 0, \beta_{i,j} \leftarrow 0.
 2 while There is still a client not covered do
       repeat
           Increase all a_j such that j is not covered yet, until next event.
4
       until An event e occurs
\mathbf{5}
       switch the event e do
 6
           case \alpha_i = c_{i,j} for some i, j do
               Call the (i, j) pair tight.
8
               Whenever \alpha_j changes in the future, set \beta_{i,j} \leftarrow \alpha_j - c_{i,j}.
               if i is already temporarily opened then Connect j to i
10
           end
11
           case \alpha_i = p_i for some client j do
12
               Choose not to connect j with penalty p_j.
13
               Call j covered but unconnected
14
           end
           case \sum_{i} \beta_{i,j} = f_i for some facility i do
16
               Temporarily open facility i.
17
               foreach uncovered or unconnected facility i with (i, j) being tight do
18
                   Connect j to i
19
               end
20
           end
21
       end
22
23 end
   // Phase 2
24 I \leftarrow any maximal set of temporarily opened facilities without conflicted.
25 foreach facility i \in I do Truely open i.
26 foreach connected client j do
       if there exists i \in I such that (i, j) is tight then truly directly connect j to i.
27
       else
28
           Let i be the facility such that j is connected to, then i is conflicted with
29
            some i' \in I since I is maximal.
           Truely indirectly connect j to i'.
30
       end
31
32 end
```

Now we analyze the cost. The cost could be decomposed into three parts:

$$X \triangleq \sum c_{i,j} x_{i,j}, \quad Y \triangleq \sum f_i y_i, \quad Z \triangleq \sum p_j z_j$$

where $x_{i,j} = 1$ if (i, j) truely connected, $y_i = 1$ if facility i is **truely opened** and $z_j = 1$ if

client j is uncovered (those we give up with penalties), and these value equals 0 otherwise.

Let I be the opened facilities, U be all the uncovered clients, C_d be the clients connected directly, C_i be the clients connected indirectly. Define $X_d = \sum_{j \in C_d} c_{i,j} x_{i,j}, X_i = \sum_{j \in C_i} c_{i,j} x_{i,j}$, then $X = X_d + X_i$.

For Z, by line 12 in algorithm 3, $\alpha_j = p_j$ for these client j, thus $Z = \sum_{j \in U} z_j = \sum_{j \in U} \alpha_j$. For $Y + X_d$,

$$Y = \sum_{i \in I} f_i = \sum_{i \in I} \beta_{i,j} = \sum_{\substack{i \in I \\ j \text{ pays for } i}} \beta_{i,j}$$

Notice that if j pays for i, then (i, j) is tight. By line 19, j is connected and thus $j \notin U$, also by line 27, $j \notin C_i$, so $j \in C_d$. Since there is no conflict in I, each client pays for at most one facility in I, hence

$$Y = \sum_{i \in I} f_i \le \sum_{\substack{i \in I, \\ j \in C_d}} \beta_{i,j}$$

And we have

$$Y + X_d = \sum_{\substack{i \in I, \\ j \in C_d}} \beta_{i,j} + \sum_{\substack{j \in C_d \\ i,j \text{ connected}}} c_{i,j} \le \sum_{j \in C_d} \alpha_j$$

Since if (i, j) is connected, (i, j) tight and thus $\alpha_j = \beta_{i,j} + c_{i,j}$.

For X_i , let j be a client in C_d , i be the client j is originally connected to and $i' \in I$ be the client j is truly indirectly connected to at the very end. Since i, i' are in conflict, there is a client j' such that j' pays for both i, i', so (i, j), (i, j'), (i', j') are all tight. j' would not be connected after j connected to i, or else once i is opened, j' would immediately be connected to i. Thus $\alpha_{j'} \leq \alpha_j$, so $c_{i',j} \leq c_{i',j'} + c_{i,j'} + c_{i,j} \leq \alpha_{j'} + \alpha_{j'} + \alpha_j \leq 3\alpha_j$ since these edges are tight. Hence

$$X_i = \sum_{\substack{j \in C_i \\ i', j \text{ connected}}} c_{i', j} \le 3 \sum_{\substack{j \in C_i }} \alpha_j$$

Combining all above, we have

$$X+Y+Z=X_i+X_d+Y+Z\leq \sum_{j\in U}\alpha_j+\sum_{j\in C_d}\alpha_j+3\sum_{j\in C_i}\alpha_j\leq 3\sum_j\alpha_j$$