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X Lessons

## Assignment: Assignment 1

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February 6, 2016

### A Primal-Dual Algorithm for Set Cover.

In this exercise, we propose to design a primal-dual algorithm for the set cover problem.

The set cover problem is as follows: given a set of elements  $E = \{e_1, \dots, e_n\}$ , some subsets of those elements  $S_1, S_2, \dots, S_m \subseteq E$ , and a non-negative weight  $w_j$  for each subset  $S_j$ . The goal is to find a minimum-weight collection of subsets that contains all the elements of  $E$ . Namely, we want to find a collection  $I$  of subsets that minimizes  $\sum_{j \in I} w_j$  and such that subject to  $\bigcup_{j \in I} S_j = E$ .

Throughout the exercise, we will consider the following linear program LP for the problem.

$$\min \sum_{j=1}^m x_j \cdot w_j$$

subject to,

$$\forall i \in \{1, \dots, n\}, \quad \sum_{j: e_i \in S_j} x_j \geq 1$$

$$\forall j \in \{1, \dots, m\}, \quad x_j \geq 0$$

Question 1: What is the dual of this linear program?

We now consider the following primal-dual algorithm.

1.  $y \leftarrow 0$

2.  $I \leftarrow \emptyset$

3. While  $I$  is not a solution (there exists  $e_i \notin I$ ):

- Increase the dual variable  $y_i$  of an element  $e_i$  that is not covered until there exists an  $l$  such that  $\sum_{j: e_j \in S_l} y_j = w_l$
- Add the set  $S_l$  to  $I$

4. Return  $I$

Correctness.

Question 2: In how many iterations of the while loop can a given dual variable be increased?

Question 3: Using Question 2, argue that the algorithm terminates and so, that the output  $I$  is a solution to the problem.

Approximation Ratio. In this section, we assume that each element of the set  $E$  appears in at most  $f$  sets of  $S_1, \dots, S_m$ .

Question 4: Recall a tight lower bound between the value of the optimal fractional solution for the dual  $\text{val}(y^*)$  and the value of the optimal integral solution for the set cover problem  $\text{OPT}$ .

Question 5: Argue that the solution  $y$  is feasible for the dual.

Question 6: Combining Questions 4 and 5, recall a tight lower bound between the value of the solution  $y$  and the value of the optimal integral solution for the set cover problem  $\text{OPT}$ .

In the following, we want to show

$$\sum_{j: S_j \in I} w_j \leq f \cdot \text{val}(y).$$

Question 7: Consider a set  $S_j \in I$ . What is the relationship between  $w_j$  and  $\sum_{i: e_i \in S_j} y_i$ ?

Question 8: Using Question 7, give the relationship between  $\sum_{j \in I} w_j$  and the variables  $y_i$ .

Question 9: Recall that  $|\{j : e_i \in S_j\}| \leq f$  for all  $i$ . Using Question 8, prove that  $\sum_{j: S_j \in I} w_j \leq f \cdot \text{val}(y)$ .

Question 10: Conclude using Questions 6 and 9.

## Answers

1. **Dual:**  $\max \sum_{i=1}^n y_i$

subject to,

$$\forall j \in \{1, \dots, m\}, \quad \sum_{i: e_i \in S_j} y_i \leq w_j$$

$$\forall i \in \{1, \dots, n\}, \quad y_i \geq 0$$

2. A given dual variable  $y_i$  corresponds to a given element  $e_i$ . If  $e_i$  is still uncovered, let's consider all the sets  $S_k \notin I$  and  $e_i \in S_k$  (There must exist at least one such set  $S_k$ . If not, then  $e_i$  can never be covered by the sets  $S_1 \dots S_m$ ). Now, in the while loop, when the variable  $y_i$  is incremented the value  $\sum_{j: e_j \in S_k} y_j$  gets incremented for all those sets  $S_k$  (since  $y_i \in S_k, \forall k$ ). But each  $S_k$  is upper-bounded by a  $w_k$  (which is finite, w.l.o.g.), so we can't go on increasing  $y_i$  indefinitely while being in the feasible region. Hence, there will be some  $l \in k$  for which the constraint  $\sum_{i: e_i \in S_k} y_i \leq w_k$  will become tight in the same iteration and it will take just one iteration for a particular element  $e_i$  after which the corresponding set  $S_l$  will be added to  $I$ .

3. Since, as argued in 2, each iteration of the while loop covers an element  $e_i$  yet to be covered (and the corresponding set covering the element is added to  $I$ ), it will take at most  $n$  iterations in the worst case to cover all the  $n$  elements in  $E$ . Hence, the while loop will terminate after at most  $n$  iterations with all the  $n$  elements covered in the solution.

4. By Weak Duality, we have  $\text{val}(y^*) \leq \text{val}(x^*)$  where  $x^*$  is the fractional optimal solution for **Primal** and  $y^*$  is the fractional optimal solution for the **Dual**. Also, the **OPT** is going to be the optimal integral solution for the **Primal minimization** problem  $\Rightarrow \text{val}(x^*) \leq \text{OPT}$ . Combining, we have  $\text{val}(y^*) \leq \text{OPT}$ .

5. Since the while loop guarantees that each of the constraints of the dual are satisfied (all the  $y_i$  variables are non-negative and the constraints are at most tight, s.t.,  $\sum_{i: e_i \in S_j} y_i \leq w_j$  is satisfied  $\forall i$ , with equality for the sets  $S_l$  that are added to  $I$  from inside the while loop). Hence the solution  $y$  remains feasible.

6. Since  $y$  is a feasible solution and  $\text{val}(y^*)$  is the optimal solution for the **Dual maximization** problem, we have  $\text{val}(y) \leq \text{val}(y^*)$ . Combining with 4 and 5, we have,  $\text{val}(y) \leq \text{OPT}$ .

7. For any set  $S_j \in I$ , we shall have  $\sum_{i: e_i \in S_j} y_i = w_j$ , as guaranteed by the while loop.

8. Hence,  $\sum_{j: S_j \in I} w_j = \sum_{j: S_j \in I} \sum_{i: e_i \in S_j} y_i = \sum_{i=1}^n |\{j : e_i \in S_j\}| y_i$ , since all the elements are there in  $I$ , with each of them probably present multiple times in multiple sets. Here  $|\{j : e_i \in S_j\}|$  present the number of times the element  $e_i$  was present in  $I$ .

9. Also, given,  $|\{j : e_i \in S_j\}| \leq f$ . Hence, we have,

$$\sum_{j: S_j \in I} w_j = \sum_{i=1}^n |\{j : e_i \in S_j\}| y_i \leq \sum_{i=1}^n f \cdot y_i = f \cdot \sum_{i=1}^n y_i = f \cdot \text{val}(y) \Rightarrow \sum_{j: S_j \in I} w_j \leq f \cdot \text{val}(y).$$

10. Combining  $\text{OPT} \geq \text{val}(y)$  (from 6) and the value of the solution provided by the **Primal-Dual algorithm**  $\leq f \cdot \text{val}(y)$  (from 9), we get the **approximation ratio** for this algorithm  $= f$ .

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