

Notes for Lecture 6

This lecture is based on Vazirani's book [1], chapters 2, 13, 14.

Outline

1. Set Cover – problem definition:

Universe U of n elements, subsets $\mathcal{S} = \{S_1, \dots, S_k\}$ with costs $c(S_i) \geq 0$.

Examples: (1) $U = a, b, c$ and $S_1 = \{a, b\}$, $S_2 = \{b, c\}$, $S_3 = \{a, c\}$, of unit cost. (2) in a museum, cover exhibits with guard positions (line of sight) (3) in a graph – cover edges with vertices or vertices with edges.

Vertex cover as a special case – each element can be covered by exactly two sets.

2. Linear programming relaxation:

$$\begin{aligned} \min \quad & \sum_{S \in \mathcal{S}} c(S)x_S \\ \text{s.t.} \quad & \sum_{S: e \in S} x_S \geq 1, \quad e \in U \end{aligned}$$

Replacing $x_S \in \{0, 1\}$ with $x_S \geq 0$.

Can be viewed as a *fractional set cover*.

Example: In ex. (1) above, can choose $x_S = 1/2$ with total cost $3/2$.

3. Relaxation provides a lower bound:

Lemma 1: The LP above can be solved in polynomial time.

Lemma 2: $\text{LP} \leq \text{OPT}$.

Importance of lower bound: Proving $\text{ALG} \leq \rho \cdot \text{LP}$ will imply $\text{ALG} \leq \rho \cdot \text{OPT}$.

4. Threshold rounding for vertex-cover:

Algorithm: Let \hat{x} be an optimal LP solution. Set $\bar{x}_S = 1$ if $\hat{x}_S \geq 1/2$, and $\bar{x}_S = 0$ otherwise.

Theorem 3: The algorithm computes a set cover of cost within factor 2 of optimal.

5. Proof:

(1) \bar{x} covers all elements, because for every element at least one of the two sets that can cover it must have $x_S \geq 1/2$.

(2) $\text{ALG} \leq 2\text{LP}$ because $c(\bar{x}) \leq 2c(\hat{x})$.

Exercise 1: Show an instance of vertex-cover where $\text{OPT} \geq 1.99\text{LP}$.

6. Randomized rounding:

Algorithm: Let \hat{x} be an optimal LP solution. Set $\bar{x}_S = 1$ independently with probability \hat{x}_S , and $\bar{x}_S = 0$ otherwise. Repeat $t = O(\log n)$ times independently and take all sets $S \in \mathcal{S}$ that were picked at least once.

Theorem 4: With probability at least $1/2$, the algorithm's output is a set cover of cost within factor $O(\log n)$ of optimal.

7. Claim: For every element $e \in U$, $\Pr[\text{a single iteration does not cover } e] \leq 1/e$.

Proof of claim: By elementary calculus, this probability is maximized when all relevant \hat{x}_S are equal, and in turn is smaller than $1/e$.

8. Proof:

(1) $\Pr[e \text{ is not covered in any iteration}] \leq (1/e)^t \leq 1/4n$, and by a union bound, all n elements are covered with probability at least $3/4$.

(2) The expected cost of one iteration is LP (by linearity of expectation). Thus, $\mathbb{E}[\text{ALG}] \leq t \cdot \text{LP}$. Using Markov's inequality, $\Pr[\text{ALG} \geq 4t \cdot \text{LP}] \leq 1/4$.

9. Exercises and Comments:

Exercise 2: Give alternate analysis by one rounding iteration where probabilities are increased by an $O(\log n)$ factor. (Hint: Chernoff)

Exercise 3: Suppose that every element has to be covered at least $5 \log n$ times, and every set can be picked any number of times. Show that a similar algorithm achieves constant approximation ratio.

Exercise 4: Show an instance where $\text{OPT} = \Theta(\log n) \cdot \text{LP}$.

Note: A greedy algorithm gives similar results but with better leading constant.

References

- [1] Vijay Vazirani. *Approximation Algorithms*. Springer, 2001. 1