高等演算法 HW1

許博翔

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Problem 1. Let OPT be the optimal solution, and val(OPT) be its value.

If $c_1 + (n-1)c_m > B$, then no worker can select the first item, and we can remove all $p_{11}, p_{21}, \ldots, p_{n1}$, so let's suppose that $c_1 + (n-1)c_m \leq B$.

Let $b = \frac{p_{11}\epsilon}{2n}$, and $p'_{ij} := \lceil \frac{p_{ij}}{b} \rceil b$, which we'll call it "new productivity".

Let
$$q_{ij} := \frac{p'_{ij}}{b} = \lceil \frac{p_{ij}}{b} \rceil$$
.

Let $dp_{ij} :=$ the minimum cost that can achieved with new productivity jb by W_1, W_2, \ldots, W_i , and r_{ij} the item that W_i should select to achieve such minimum cost. The range: $1 \le i \le n$, $0 \le j \le Q$, $Q := \sum_{k=1}^{n} q_{k1}$.

The base case i = 1:

$$dp_{1j} = \begin{cases} \min_{k:q_{1k}=j} (c_k), & \text{if } \exists k \text{ s.t. } q_{1k}=j\\ B+1, & \text{otherwise} \end{cases}$$

$$r_{1j} = \begin{cases} k, \text{ where } c_k = dp_{1j} \text{ and } q_{1k} = j, \text{ if such } k \text{ exists} \\ -1, \text{ otherwise} \end{cases}$$

One can run from i=2 to n, from j=0 to Q to get the values of dp_{ij} using

$$dp_{ij} = \begin{cases} \min_{k:q_{ik} \le j} (dp_{i-1,j-q_{ik}} + c_k), & \text{if } \exists k \text{ s.t. } q_{ik} \le j \\ B+1, & \text{otherwise} \end{cases}$$

$$r_{ij} = \begin{cases} k, \text{ where } dp_{i-1,j-q_{ik}} + c_k = dp_{ij} \text{ and } q_{ik} \leq j, \text{ if such } k \text{ exists} \\ -1, \text{ otherwise} \end{cases}$$

Denote the optimal solution as ALG, and the value of ALG (denote as val'(ALG) is the maximum new productivity that can be achieved with cost at most B, which

is $\max_{j:dp_{nj}\leq B}(jb)$, and we can recursively find the selected item that can achieve this using r_{ij} .

The above can be done in O(nQm) time complexity.

Let j_i denote the selected item by W_i in ALG, and let $\sum p_{ij_i}$ be the productivity value of this algorithm (denoted as val(ALG)).

Let k_i denote the selected item by W_i in OPT, and let the new productivity value of these selected item be val'(OPT).

By the definition of OPT, $val(ALG) \leq val(OPT)$.

Since the above dp algorithm obtains optimal solution of new productivity, val'(ALG) > 0val'(OPT).

$$val(ALG) = \sum_{i=1}^{n} p_{ij_i} > \sum_{i=1}^{n} (\lceil \frac{p_{ij_i}}{b} \rceil - 1)b = val'(ALG) - nb \ge val'(OPT) - nb = val'(ALG)$$

$$\sum_{i=1}^{n} \lceil \frac{p_{ik_i}}{b} \rceil b - nb \ge \sum_{i=1}^{n} p_{ik_i} - nb = val(OPT) - nb = val(OPT) - \frac{p_{11}\epsilon}{2}.$$

By what we suppose in the first three lines, $p_{11} \leq p_{11} + p_{2m} + p_{3m} + \cdots + p_{nm} \leq$

$$val(OPT)$$
 (since W_1 can select 1, while W_2, W_3, \dots, W_n select m).

$$\Rightarrow val(ALG) > val(OPT) - \frac{val(OPT)\epsilon}{2} = (1 - \frac{\epsilon}{2})val(OPT)$$

$$\Rightarrow val(ALG) \ge val(OPT) - \frac{val(OPT)\epsilon}{2} = (1 - \frac{\epsilon}{2})val(OPT).$$

$$\Rightarrow val(ALG) \le val(OPT) \le \frac{val(ALG)}{1 - \frac{\epsilon}{2}} \le (1 + \epsilon)val(ALG).$$

The time complexity of this algorithm = $O(nQm) = O(n\sum_{k=1}^{n}q_{k1}m) = O(n\sum_{k=1}^{n}\lceil\frac{p_{k1}}{b}\rceil m) =$

$$O(n\sum_{k=1}^{n} \lceil \frac{2np_{k1}}{p_{11}\epsilon} \rceil m) = O(n\sum_{k=1}^{n} \frac{2n}{\epsilon} m) = O(\frac{n^3m}{\epsilon})$$

Problem 2. Let OPT be the optimal solution, and val(OPT) be its value.

Let
$$p_j := p_{1j} = p_{2j} = \dots = p_{nj}$$
.

First, there is a 4-approximation.

Let
$$k = \min_{i:p_{p_{A_{i+1}}+p_{A_{i+2}}+p_{A_{i+3}}+p_{A_{i+4}} \le P} (i)$$

Let $k = \min_{i: p_{p_{4i+1} + p_{4i+2} + p_{4i+3} + p_{4i+4} < P} (i)$. That is, for $i = 1, 2, \dots, k, p_{4i+1} + p_{4i+2} + p_{4i+3} + p_{4i+4} \ge P$.

$$\Rightarrow k \leq val(OPT)$$
.

Since $p_{4k+1} \geq p_{4k+2} \geq \cdots \geq p_m$, for all 4 distinct elements a, b, c, d of the multiset

$${p_{4k+1}, p_{4k+2}, \dots, p_m}, a+b+c+d \le p_{4i+1} + p_{4i+2} + p_{4i+3} + p_{4i+4} < P.$$

 \Rightarrow a worker with productivity at least P should take at least one of the $1, 2, \ldots, 4k$ th machine.

$$\Rightarrow val(OPT) \leq 4k.$$

$$\therefore k \le val(OPT) \le 4k.$$

Since in the OPT solution, one would use at most 16k machines, and using the machines with larger productivity will not decrease the number of workers with productivity at least P.

 \therefore set $M := \min(16k, m)$, and there is an OPT solution s.t. only the *i*-th $(1 \le i \le M)$ machine will be used.

Let
$$a := \lfloor \frac{M\epsilon}{32} \rfloor$$
, and $b := \lceil \frac{M}{a} \rceil$.

Let $q_i := p_{M+1-i}$. (That is, q is p's reverse, which is increasing.)

Partition $\{q_1, q_2, ..., q_M\}$ into $S_1, S_2, ..., S_b$, where $S_i := \{q_j : a(i-1) + 1 \le j \le ai\}$.

That is, the *i*-th machine is of the $\lceil \frac{i}{a} \rceil$ -th type, and define the new productivity of the machines of the *j*-th type as $f(S_i)$.

There are at most $c := \begin{pmatrix} b \\ 4 \end{pmatrix} + \begin{pmatrix} b \\ 3 \end{pmatrix} + \begin{pmatrix} b \\ 2 \end{pmatrix} + \begin{pmatrix} b \\ 1 \end{pmatrix} + \begin{pmatrix} b \\ 0 \end{pmatrix}$ ways to select the types of at most 4 different machines.

There are n identical workers in total, and c different ways to select the types of the machines they take.

 \Rightarrow there are at most $\binom{c}{n}$ possibilities.

Bruteforce through all (at most) $\binom{c}{n}$ possibilities, for each possibility, check if the *i*-th type of machine is used by at most $|S_i|$ workers for $i=1,2,\ldots,b$, and then calculate the number of workers with new productivity ≥ 4 . The value of this algorithm with new productivity function f (denote as val(f)) is the maximum number of workers with new productivity ≥ 4 . The complexity of this part is $O(\binom{c}{n}) \times (b+n) = O(n^{c+1})$.

Define
$$f_1$$
: $f_1(S_i) := \begin{cases} \min(S_i), & \text{if } i \geq 2 \\ 0, & \text{if } i = 1 \end{cases}$
Define f_2 : $f_2(S_i) := \begin{cases} \min(S_{i+1}), & \text{if } i \leq b-1 \\ \max(P, q_M), & \text{if } i = b \end{cases}$

Since the difference of the new productivities using f_1 , f_2 are a 0s, and a max (P, q_M) , and the a worker taking only max (P, q_M) have new productivities $\geq P$.

$$\therefore val(f_2) \le val(f_1) + a.$$

Also, the new productivity of the *i*-th machine in f_1 is not larger than the original productivity, and in f_2 is not smaller than the original productivity.

Problem 3.

Problem 4. Let w_v be the weight of the vertex $v \in V(G)$.

Transform the vertex cover problem to an ILP problem (like that taught in class):

Variables: $\{x_v : v \in V(G)\}.$

$$\min \sum_{v \in V} w_v x_v.$$

Subject to:

 $x_v \in \{0,1\}, \ \forall v \in V(G), \text{ where } x_v = 1 \text{ iff the vertex cover contains } v.$

$$x_u + x_v \ge 1, \ \forall uv \in E(G).$$

Relax the above to LP (that is, relax the condition $x_v \in \{0, 1\}$ to $0 \le x_v \le 1$), then we can solve it in polynomial time complexity, and suppose the solution is $x_v = y_v^*$. Let $I \subseteq V(G)$ be an independent set.

One can see that
$$y_v^{(I)} := \begin{cases} 0, & \text{if } y_v^* < \frac{1}{2} \text{ or } (x_v = \frac{1}{2} \text{ and } v \in I) \\ 1, & \text{otherwise} \end{cases}$$
 satisfy that $y_v^{(I)} \in \{0, \text{ if } y_v^* < \frac{1}{2} \text{ or } (x_v = \frac{1}{2} \text{ and } v \in I) \end{cases}$

$$\{0,1\}, \ \forall v \in V(G).$$

Since $y_u^* + y_v^* \ge 1$, WLOG suppose that $y_u^* \ge y_v^*$, there is $y_u^* \ge \frac{1}{2}y_u^* + y_v^* \ge \frac{1}{2}$.

If
$$y_u^* > \frac{1}{2}$$
 or $(y_u^* = \frac{1}{2} \text{ and } u \notin I)$, then $y_u^{(I)} = 1$.

Otherwise, $y_u^* = \frac{1}{2}$ and $u \in I$.

$$\Rightarrow y_v^* \ge 1 - y_u^* = \frac{1}{2}.$$

Since I is an independent set and $uv \in E$ and $u \in I$, there must be $v \notin I$.

$$\Rightarrow y_v^{(I)} = 1.$$

 \therefore at least one of $y_u^{(I)}, y_v^{(I)} = 1$.

 \Rightarrow the condition " $y_u^{(I)} + x_v^{(I)} \ge 1$, $\forall uv \in E(G)$ " is satisfied.

In class, we learn that the solution to this LP problem satisfies $\forall v \in V(G), \ y_v^* \in \{0, \frac{1}{2}, 1\}.$

Since G has a k-coloring, one can partition V(G) into k independent sets I_1, I_2, \dots, I_k .

If
$$y_v^* = \frac{1}{2}$$
, then $\sum_{v \in V(G)} y_v^{(I_i)} = 1(k-1) + 0 = k-1 = (2k-2)y_v^*$.

If
$$y_v^* = 0$$
 or 1, then $\sum_{v \in V(G)} y_v^{(I_i)} = k y_v^* \le (2k - 2) y_v^*$.

$$\therefore \sum_{i=1}^{k} \sum_{v \in V(G)} y_v^{(I_i)} \le \sum_{v \in V(G)} (2k-2) y_v^*.$$

By pigeonhole principle, $\exists i \text{ s.t. } \sum_{v \in V(G)} y_v^{(I_i)} \leq (2 - \frac{2}{k}) \sum_{v \in V} y_v^*.$

Let val(OPT) be the value of the ILP.

Since LP relaxes some condition of the ILP, $\sum_{v \in V} y_v^* \leq val(OPT)$.

$$\Rightarrow \sum_{v \in V(G)} y_v^{(I_i)} \le (2 - \frac{2}{k}) \sum_{v \in V} y_v^* \le (2 - \frac{2}{k}) val(OPT).$$

The time complexity is polynomial since:

- 1. The time complexity creating and solving the LP problem is polynomial.
- 2. The time complexity running through all i=1 to k, finding the i such that $\sum_{v\in V(G)}y_v^{(I_i)}\leq (2-\frac{2}{k})\sum_{v\in V(G)}y_v^* \text{ is } O(kV(G))\overset{k\leq V(G)}{\leq} O(V(G)^2).$