## 22C:253 Algorithms for Discrete Optimization

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October 21, 2002

We will quickly wrap up our discussion of the problem Scheduling on Unrelated Parallel Machines (SUPM) that started last lecture. We now have a solution  $(T^*, X^*)$  to LP + (C). We showed that at most m jobs are assigned fractionally in  $x^*$ . There is a simple way in which these fractional jobs can be rounded.

Consider the bipartite graph G = (A, B, E) such that A is the set of machines which are assigned some fractional jobs, B is the set of jobs assigned fractionally, and E contains edges  $\{i, j\}$ , where  $i \in A$ ,  $j \in B$  and  $x_{ij} \in (0, 1)$ . This is a bipartite graph and it can be shown that this contains a matching in which all jobs are matched. This is left as an exercise for you.

As mentioned in the last lecture, in order to "round"  $x^*$  given the above property, we simply assign each job to the machine it is matched with. This increases the makespan from  $T^*$  to at most  $2 \cdot T^* \leq 2 \cdot OPT$ .

Family of Tight Examples. Let  $n = m^2 - m + 1$  where n is the number of jobs and m denote the number of machines. Suppose job-1,  $j_1$ , has a processing time m on all machines and any other job,  $j_i$ , can be processed in unit time on any machine.

The *OPT* for this problem instance is m. Say,  $j_1$  is assigned to  $m_1$  and completes in time m. The remaining  $m^2 - m$  jobs are assigned so that each of the remaining m - 1 machines get m jobs. In fact, the above solution is a feasible solution for the LP. Let us consider the following feasible solution.

- Split  $j_1$  into unit sized jobs and assign one unit to each machine.
- Of the remaining jobs, assign (m-1) of these to each machine.

This solution is a vertex of the feasibility polytope and forms a feasible solution with makespan m. If this solution is returned by the LP relaxation, then rounding will assign  $j_1$  to one of the machines and increase the makespan to 2m - 1.

## CAPACITATED VERTEX COVER (CapVC)

INPUT: Let G = (V, E) is a graph with vertex weights  $w_v \in Q^+$  and vertex capacities  $k_v \in Z^+$ . OUTPUT: A vertex cover defined by a function,  $x : V \to N_0$  such that

- (i) There is an orientation of the edges such that the number of edges coming into any vertex is at most  $k_v \cdot x(v)$ .
- (ii)  $\sum_{v \in V} w_v \cdot x(v)$  is minimized.

Status of the Problem: A factor-2 approximation can be obtained by using dependent rounding and an alternate factor-2 approximation algorithm can be obtained using the primal-dual framework. We will discuss a simple factor-4 algorithm that uses a deterministic rounding technique and a factor-3 approximation algorithm using dependent rounding method. Given below is the integer program corresponding to CapVC. The variables used are:  $x_v \in N_0$  for each  $v \in V$  and  $y_{e,v} \in \{0, 1\}$ for each edge  $e \in E$  and  $v \in e$ .  $y_{e,v}$  indicates if vertex v covers e.

$$\text{minimize} \sum_{v \in V} w_v \cdot x_v$$

such that

$$\begin{array}{rcl} y_{e,v} + y_{e,u} & \geq & 1 & \text{for each edge} & e = \{u, v\} \in E \\ & \sum_{e: v \in e} y_{e,v} & \leq & k_v \cdot x_v & \text{for each} & v \in V \\ & & x_v & \in & N_0 \\ & & y_{e,v} & \in & \{0, 1\} \end{array}$$

The corresponding LP relaxation replaces the constraints  $y_{e,v} \in \{0,1\}$  by  $y_{e,v} \ge 0$  and  $x_v \in N_0$  by  $x_v \ge 0$ . Any feasible solution to the above IP satisfies the following property:

If  $y_{e,v} = 1$  for some edge  $e : v \in e$ , then  $x_v \ge 1$ .

This property can be enforced in the LP relaxation problem by adding the following linear constraint

$$x_v \ge y_{e,v}$$
 for each  $e: v \in e$ 

**Deterministic Rounding Algorithm.** Here is a deterministic rounding algorithm that yields a factor-4 approximation.

- 1. Solve the LP-relaxation to obtain the solution (X, Y).
- 2. For each  $y_{e,v} \ge \frac{1}{2}$ ,  $y_{e,v}^{\star} = 1$ . For all other  $y_{e,v}$ , set  $y_{e,v}^{\star} = 0$ .
- 3. Set

$$x_v^{\star} = \lceil \frac{\sum_{e:v \in e} y_{e,v}^{\star}}{k_v} \rceil \tag{1}$$

<u>Claim</u>: This algorithm produces a factor-4 approximation algorithm.

**Proof:** We know that  $y_{e,v}^{\star} \leq 2y_{e,v} \ \forall e \in E, v \in e$ . And we want to show that:  $x_v^{\star} \leq 4x_v \quad \forall v \in V$ . Since,

$$y_{e,v}^{\star} \le 2 \cdot y_{e,v}$$

we get,

$$y_v^{\star} = \sum_{e:v \in e} y_{e,v}^{\star} \leq 2 \cdot \sum_{e:v \in e} y_{e,v} \leq 2 \cdot k_v \cdot x_v \tag{2}$$

Let,

$$y_v^{\star} = ak_v + b \quad \forall \ a, b \in I, a \ge 0, \quad 0 \le b \le k_v.$$

$$\tag{3}$$

So,

$$x_v \ge \frac{ak_v + b}{2k_v} = \frac{a}{2} + \frac{b}{2k_v} \tag{4}$$

Now using (1) and (3),

$$x_v^{\star} = \lceil \frac{y_v^{\star}}{k_v} \rceil \quad \le \quad a+1 \tag{5}$$

Therefore, if we can show that

$$(a+1) \le 4(\frac{a}{2} + \frac{b}{2k_v}) \le 2a + \frac{2b}{k_v}$$

we will be done.

Now,  $RHS = 2a + \frac{2b}{k_v}$  If  $a \ge 1$  then  $RHS \ge LHS$ . If a = 0, then  $y_v^* \le k_v$ . This implies that  $x_v^* \in \{0, 1\}$ . If  $x_v^* = 0$ , then we are done. If  $x_v^* = 1$ , then  $y_v^* = 1$  Hence,  $y_{e,v}^* = 1$  for some edge  $e : v \in e$ . This implies  $y_{e,v} \ge \frac{1}{2}$  for some  $e : v \in e$ . Therefore,  $x_v \ge \frac{1}{2}$  by the constraint added to the LP relaxation. Hence,  $x_v^* \le 4x_v$ .  $\Box$