Approximation Algorithms Lecture 13

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1 Recap

APTAS for bin packing.

2 Content

2.1 Linear program for vertex cover

Our template would be the following: make an integer program, do a suitable relaxation to get a linear program, then solve it to get an optimal solution of the LP, then round this fractional solution to get an integer solution, which could be a solution to the original problem too.

So for vertex cover, we have a graph G = (V, E), and we want to pick a set of vertices, which covers all edges.

Let $x_v = 1$ if $v \in V$ is in the vertex cover, and 0 otherwise.

What should the constraints on these variables be to ensure that the vertices v with $x_v = 1$ form a vertex cover?

$$\forall e = (u, v) \in E : x_u + x_v \ge 1.$$

The objective function is $\sum_{v \in V} x_v$.

So the integer program becomes something like the following:

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

$$\forall e = (u, v) \in E : x_u + x_v \ge 1$$

$$\forall v \in V : x_v \in \{0, 1\}$$

Note that the integer program always corresponds to a vertex cover.

Relaxing this to a linear program gives us:

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

$$\forall e = (u, v) \in E : x_u + x_v \ge 1$$

$$\forall v \in V : 0 \le x_v \le 1$$

Claim 0.1

The constraint $x_v \leq 1$ is redundant.

Proof. Replace all $x_v > 1$ by $x_v = 1$, and note that it doesn't violate any constraint, but only decreases the value of the objective function.

Hence, the linear program is equivalent to

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

$$\forall e = (u, v) \in E : x_u + x_v \ge 1$$

$$\forall v \in V : x_v \ge 0$$

Let's try to write the dual program of this.

Note 1

Note that the canonical forms are the following:

$$\max c^{\top} x$$
$$Ax \le b$$
$$x \ge 0$$

and

$$\min c^{\top} x$$
$$Ax \ge b$$
$$x \ge 0$$

To get the multipliers, we do the following: the objective function has coefficient of u being 1.

So the dual is:

$$\min \sum_{e \in E} y_e$$

$$\forall u \in V : \sum_{e \in \delta(u)} y_e \le 1$$

$$\forall e \in E : y_e \ge 0$$

where $\delta(u)$ is the set of edges incident to a vertex u.

Suppose primal is a minimization problem, then dual is a maximization problem.

So the value of the dual will always be on the left of the value of the primal.

Now let's come back to the primal.

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

$$\forall e = (u, v) \in E : x_u + x_v \ge 1$$

$$\forall v \in V : x_v \ge 0$$

Let x^* be the optimal solution to this linear program.

Then we have $\sum_{v \in V} x_v^* \leq \mathbf{OPT}_{VC}$, since we have only relaxed the constraints. Note that x^* is not integral, i.e., it could be fractional.

(For bipartite, x^* is integral, and in general it is half-integral, and the dual corresponds to the maximum independent set).

We round x^* to an integral solution as follows: if $x*_v \ge 1/2$, round it to 1, else to 0.

Is \overline{x} a feasible solution to the integer program? Yes, since if $\overline{x}_v = 0$ and $\overline{x}_u = 0$, then $x_v^* < \frac{1}{2}$ and $x_u^* < \frac{1}{2}$, which would violate the inequality.

Now let's look at the quality of the solution. We have the size of the vertex cover as:

$$\sum_{v \in V} \overline{x}_v \le \sum_{v \in V} 2x_v^*$$

$$\le 2 \cdot \mathbf{OPT}_{VC}$$

This gives us a 2-approximation.

Now suppose we have weights on the vertices, i.e., a function $w:V\to\mathbb{R}^+$, and we want a minimum weight vertex cover.

We need to just change the objective function by replacing x_v by $w(v)x_v$, and the rest of the proof holds.

Let's look at what the dual program becomes. We just need to change the RHS from 1 to w(u).

2.2 Linear program for set cover

Let
$$U = \{e_1, \dots, e_n\}$$
, and $S_1, \dots, S_m \subseteq U$.

 x_i is the variable corresponding to S_i such that $x_i = 1$ if S_i is in the set cover, and 0 otherwise.

Now we need: $\forall e_j \in U : \sum_{i:e_j \in S_i} x_i \ge 1$, and $x_i \in \{0,1\} \forall i$.

Our objective function is $\sum_i x_i$.

The corresponding integer program is:

$$\min \sum_{i} x_{i}$$

$$\forall e_{j} \in U : \sum_{i: e_{j} \in S_{i}} x_{i} \geq 1$$

$$x_{i} \in \{0, 1\} \forall i$$

By the same argument as before, we can relax it to:

$$\min \sum_i x_i$$

$$\forall e_j \in U : \sum_{i: e_j \in S_i} x_i \ge 1$$

$$x_i \ge 0 \forall i$$