Linear Programming and Primal-Dual Schema

Chihao Zhang

BASICS, Shanghai Jiao Tong University

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Example: Set Cover

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Input: A Universe E = \{e_1, \dots, e_n\}; a family of sub-
         sets S_1, \ldots, S_m where each S_i \subseteq E; a nonnegative
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weight $w_i \geq 0$ for each S_i .

Problem: Find a minimum-weight collection of subsets that

covers all of E

Integer Program

minimize
$$\sum_{j=1}^m w_j x_j$$
 subject to $\sum_{j:e_i \in S_j} x_j \geq 1, \quad i=1,\ldots,n,$ $x_j \in \{0,1\}, \quad j=1,\ldots,m.$

 $x_i \in \{0,1\}$: indicate whether S_i is in the solution.

Linear Program Relaxation

minimize
$$\sum\limits_{j=1}^m w_j x_j$$
 subject to $\sum\limits_{j:e_i \in S_j} x_j \geq 1, \quad i=1,\ldots,n,$ $x_j \geq 0, \qquad j=1,\ldots,m.$

maximize
$$\mathbf{c}^T \mathbf{x}$$

subject to
$$A \mathbf{x} \leq \mathbf{b}$$

$$\mathbf{x} > 0$$

$$A = (a_{i,j})$$
: An $m \times n$ matrix
 $\mathbf{b} = (b_1, \dots, b_m)$: A vector of m entries
 $\mathbf{c} = (c_1, \dots, c_n)$: A vector of n entries

Every LP can be transformed to canonical form efficiently.

- Simplex Algorithm
- Ellipsoid Method

Dual of Linear Program

Consider the following linear program:

```
maximize x_1 + 6x_2

x_1 \le 200 (1)

x_2 \le 300 (2)

x_1 + x_2 \le 400 (3)

x_1, x_2 \ge 0
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The optimal solution is at $(x_1, x_2) = (100, 300)$, with objective value 1900.

Dual of Linear Program (cont'd)

$$(1) + 6 \times (2) : x_1 + 6x_2 \le 2000.$$

 $0 \times (1) + 5 \times (2) + 1 \times (3) : x_1 + 6x_2 \le 1900$

Multiplier	Inequality				
<i>y</i> ₁	x_1			\leq	200
<i>y</i> ₂			<i>x</i> ₂	\leq	300
<i>y</i> 3	<i>X</i> ₁	+	X2	<	400

Dual of Linear Program (cont'd)

minimize
$$200y_1 + 300y_2 + 400y_3$$

 $y_1 + y_3 \ge 1$
 $y_2 + y_3 \ge 6$
 $y_1, y_2, y_3 \ge 0$

Dual of Linear Program (cont'd)

Primal IP:

maximize
$$\mathbf{c}^T \mathbf{x}$$

$$A\mathbf{x} \leq \mathbf{b}$$

$$\mathbf{x} \geq 0$$

Dual LP:

minimize
$$\mathbf{y}^T \mathbf{b}$$

$$\mathbf{y}^T A \ge \mathbf{c}^T$$

$$\mathbf{y} \ge 0$$

weakstrong duality property:

$$\mathbf{c}^T\mathbf{x} \leq \mathbf{y}^T\mathbf{b}\mathbf{c}^T\mathbf{x}^* = (\mathbf{y}^*)^T\mathbf{b}$$

A Simple Rounding Algorithm for Set Cover

Recall the linear programming relaxation for set cover:

minimize
$$\sum_{j=1}^m w_j x_j$$
 subject to $\sum_{j:e_i \in S_j} x_j \geq 1, \quad i=1,\ldots,n,$ $x_j \geq 0, \qquad \qquad j=1,\ldots,m.$

Consider the optimal solution of the LP. Intuitively, the set with larger x_j is more likely to be in a good solution of set cover (or the IP).

A Simple Rounding Algorithm for Set Cover (cont'd)

- 1. Let x* be the solution of LP.
- 2. Let $I = \{j \mid x_j^* \ge 1/f\}$, where $f = \max_{i=1,...,n} |\{j \mid e_i \in S_j\}|$.
- 3. Output 1.

Lemma

S is a set cover.

Analysis

Lemma

The rounding algorithm is an f-approximation algorithm for the set cover problem.

$$\sum_{j \in I} w_j \le \sum_{j=1}^m w_j \cdot (f \cdot x_j^*)$$

$$= f \sum_{j=1}^m w_j x_j^*$$

$$= f \cdot \text{OPT}$$

Recall the LP relaxation for Set Cover:

minimize
$$\sum_{j=1}^m w_j x_j$$
 subject to $\sum_{j:e_i \in S_j} x_j \geq 1, \quad i=1,\ldots,n,$ $x_j \geq 0, \qquad \qquad j=1,\ldots,m.$

and its dual:

maximize
$$\sum_{i=1}^n y_i$$

subject to $\sum_{i:e_i \in S_j} y_i \le w_j, \quad j=1,\ldots,m,$
 $y_i > 0, \qquad \qquad i=1,\ldots,n.$

Vertex Cover

- Vertex cover problem is the special case of set cover problem when f = 2.
- The dual of vertex cover problem is maximum matching problem.
- The duality theorem implies

maximum matching < minimum vertex cover

Vertex Cover (cont'd)

Consider the following algorithm:

- 1. $M \leftarrow \emptyset$
- 2. $S \leftarrow \emptyset$
- 3. while G is not empty do
 - 3.1 Choose an edge $e = \{u, v\} \in E(G)$ and let $M \leftarrow M \cup \{e\}$
 - 3.2 $S \leftarrow S \cup \{u, v\}$
 - 3.3 $G \leftarrow G[V \setminus \{u, v\}]$ (Remove isolated nodes)
- 4. return S

This is the combinatorial interpretation of a primal-dual algorithm.

The Algorithm

- 1. $\mathbf{y} \leftarrow 0$ 2. $\mathbf{I} \leftarrow \emptyset$
- 3. **while** there exists $e_i \notin \bigcup S_j$ **do**
 - 3.1 Increase the dual variable y_i until there is some ℓ such that $\sum y_j = w_\ell$ 3.2 $I \leftarrow I \cup \{\ell\}$
- 4. return /.

Analysis

The primal-dual algorithm is an f-approximation algorithm for the set cover problem.

$$\sum_{j \in I} w_j = \sum_{j \in I} \sum_{i: e_i \in S_j} y_i$$

$$= \sum_{i=1}^n y_i \cdot |\{j \in I \mid e_i \in S_j\}|$$

$$\leq f \cdot \text{OPT}$$

Complementary Slackness

The following property is called complementary slackness.

$$\sum_{i=1}^{n} y_{i} \leq \sum_{i=1}^{n} y_{i} \sum_{j:e_{i} \in S_{j}} x_{j} = \sum_{j=1}^{m} x_{j} \sum_{i:e_{i} \in S_{j}} y_{i} \leq \sum_{j=1}^{m} x_{j} w_{j}.$$

Let x^* and y^* be the optimal solution of primal and dual LP respectively, then

•
$$y_i^* > 0 \implies \sum_{j:e_i \in S_j} x_j^* = 1$$
,

•
$$x_j^* > 0 \implies \sum_{i:e_i \in S_i} y_i^* = w_j$$
.

$$\sum_{j \in I} w_j = \sum_{j \in I} \sum_{i: e_i \in S_j} y_i$$

$$= \sum_{i=1}^n y_i \cdot |\{j \in I \mid e_i \in S_j\}|$$

$$= \sum_{i=1}^n y_i \cdot \sum_{j: e_i \in S_j} x_j$$

$$< f \cdot \text{OPT}$$

where $x_i \in \{0,1\}$ and $x_j = 1$ if and only if $j \in I$.

Discussion

The primal-dual algorithm ensures

$$x_j > 0 \implies \sum_{i: e_i \in S_j} y_i = w_j$$

In general, we cannot hope

$$y_i > 0 \implies \sum_{j: e_i \in S_i} x_j = 1$$

We want to show it is not too slack, i.e.

$$y_i > 0 \implies \sum_{j:e_i \in S_i} x_j \leq \alpha$$

Feedback Vertex Set Problem

Input: A undirected graph G = (V, E) and nonnegative

weights $w_i > 0$ for $i \in V$.

Problem: Find a set $S \subseteq V$ of minimum weight such that

 $G[V \setminus S]$ is a forest

LP formulation

minimize
$$\sum_{i \in V} w_i x_i$$
 subject to $\sum_{i \in C} x_i \geq 1$, $orall C \in \mathcal{C}$ $x_i \in \{0,1\}, \ orall i \in V$

minimize
$$\sum_{i \in V} w_i x_i$$

subject to $\sum_{i \in C} x_i \ge 1$, $\forall C \in \mathcal{C}$
 $x_i > 0$, $\forall i \in V$

Dual LP

maximize
$$\sum_{C \in \mathcal{C}} y_C$$
 subject to $\sum_{C \in \mathcal{C}: i \in C} y_C \leq w_i, \ \forall i \in V,$ $y_C \geq 0, \ \forall C \in \mathcal{C}$

The Algorithm

- 1. $\mathbf{y} \leftarrow 0$
- 2. $S \leftarrow \emptyset$
- 3. **while** there exists a cycle *C* in *G* **do**
 - 3.1 Increase y_C until there is some $\ell \in V$ such that

$$\sum_{C' \in \mathcal{C}: \ell \in C'} y_{C'} = w_{\ell}$$

- 3.2 $S \leftarrow S \cup \{\ell\}$
- 3.3 Remove ℓ from G
- 3.4 Repeatedly remove vertices of degree one from G
- 4. return S.

Analysis

$$\sum_{i\in S} w_i = \sum_{i\in S} \sum_{C:i\in C} y_C = \sum_{C\in C} |S\cap C| y_C.$$

 $|S \cap C|$ may be as large as |V|!

Observation

For any path P of vertices of degree two in graph G, our algorithm will choose at most one vertex from P.

Theorem

In any graph G that has no vertices of degree one, there is a cycle with at most $2\lfloor \log_2 n \rfloor$ vertices of degree three or more, and it can be found in linear time.

Algorithm (revised)

- 1. $\mathbf{y} \leftarrow 0$
- 2. $S \leftarrow \emptyset$
- 3. Repeatedly remove vertices of degree one from G
- 4. **while** there exists a cycle C in G **do**
 - 4.1 Find cycle C with at most $2|\log_2 n|$ vertices of degree three or more
 - 4.2 Increase y_C until there is some $\ell \in V$ such that

$$\sum_{C' \in \mathcal{C}: \ell \in C'} y_{C'} = w_{\ell}$$

- 4.3 $S \leftarrow S \cup \{\ell\}$
- 4.4 Remove & from G
- 4.5 Repeatedly remove vertices of degree one from G
- 5. return S.

Analysis

$$\sum_{i \in S} w_i = \sum_{C \in \mathcal{C}} |S \cap C| y_C \le (4 \lfloor \log_2 n \rfloor) \sum_{C \in \mathcal{C}} y_C \le (4 \lfloor \log_2 n \rfloor) \text{OPT}.$$