

Primal-Dual Schema

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Motivation So far we have been talking about linear programs and it's applications to finding approximate solutions for various integer programs. This lecture introduces a method for analyzing these types of approximations. First a quick review:

Linear Programs A linear program is given by a tuple (A, b, c) where A is an $n \times m$ matrix, b is a $n \times 1$ column vector, and c a $1 \times m$ row vector and defines the following (potentially empty) convex set:

$$\operatorname{argmin}_{x \in \mathbb{R}^m \mid Ax \leq b} cx$$

The dual of a linear program is defined as

$$(A, b, c)^* = (-A^T, -c^T, -b^T)$$

Which gives the set

$$\operatorname{argmin}_{y \in \mathbb{R}^n \mid -A^T y \leq -c^T} -b^T y = \operatorname{argmax}_{y \in \mathbb{R}^n \mid A^T y \geq c^T} b^T y$$

A solution to a program is a point within the aforementioned set. The set $\{x \mid Ax \leq b\}$ is known as the feasible space of a linear program. Recall that if a pair of points x, y are a solution for $(A, b, c), (A, b, c)^*$ respectively if and only if

$$\forall j \in [1, m] : x_j \sum_{i=1}^n A_{i,j} y_i = c_j x_j \quad (1)$$

$$\forall i \in [1, n] : y_i \sum_{j=1}^m A_{i,j} x_j = b_i y_i \quad (2)$$

Approximate Complementary Slackness Equations 1 and 2 are called the primal and dual complementary slackness conditions and are both necessary and sufficient conditions for optimality. We now introduce a somewhat less strict condition on a (possibly infeasible) solution pair (x, y) for (A, b, c) known as approximate slackness; given scalars $\alpha, \beta \geq 1$

$$\forall j \in [1, m] : \frac{c_j}{\alpha} x_j \leq x_j \sum_{i=1}^n A_{i,j} y_i \leq c_j x_j \quad (3)$$

$$\forall i \in [1, n] : b_i y_i \leq y_i \sum_{j=1}^m A_{i,j} x_j \leq \beta b_i y_i \quad (4)$$

Now suppose that equations 3, 4 hold for some (x, y) ; then we know:

$$\begin{aligned}
\sum_{j=1}^m \frac{c_j}{\alpha} x_j &\leq \sum_{j=1}^m x_j \sum_{i=1}^n A_{i,j} y_i \\
\sum_{j=1}^m x_j \sum_{i=1}^n A_{i,j} y_i &= \sum_{i=1}^n y_i \sum_{j=1}^m A_{i,j} x_j \\
\sum_{i=1}^n y_i \sum_{j=1}^m A_{i,j} x_j &\leq \sum_{i=1}^n \beta b_i y_i \\
\sum_{j=1}^m \frac{c_j}{\alpha} x_j &\leq \sum_{i=1}^n \beta b_i y_i
\end{aligned}$$

And thus we arrive at the following condition:

$$\sum_{j=1}^m c_j x_j \leq \alpha \beta \sum_{i=1}^n b_i y_i \tag{5}$$

The values α, β determine the amount of deviation from an optimality. In the case where $\alpha = \beta = 1$, then the exact complementary slackness of equations 1, 2 is satisfied and we would have an optimal solution. As α, β grow larger, the bounds on the quality of the approximation gets worse.

Example: Set Cover The set cover problem can be stated as follows: given a collection of sets \mathcal{S} , some points $U \subset \bigcup_{S \in \mathcal{S}} S$ and a scalar cost function $c(S) : \mathcal{S} \rightarrow \mathbb{R}^+$, find the collection of sets $X \subset \mathcal{S}$ such that $U \subseteq \bigcup_{S \in X} S$ minimizing $\sum_{S \in X} c(S)$. This problem can be concisely written as an integer program

$$\begin{array}{ll}
\text{Primal:} & \underset{x \in \mathbb{Z}^{\mathcal{S}} | e \in U \rightarrow \sum_{e \in S} x_S \geq 1}{\operatorname{argmin}} \quad c(S)x_S \\
\text{Dual:} & \underset{y \in \mathbb{R}^U | S \in \mathcal{S} \rightarrow \sum_{e \in S} y_e \leq c(S)}{\operatorname{argmax}} \quad y_e
\end{array}$$

Now consider the following approximation algorithm for set-cover:

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1   $x \leftarrow 0, y \leftarrow 0$ 
2  while  $\exists y_e = 0$ 
3      do Pick an uncovered  $e$ 
4           $S_{min} \leftarrow \emptyset, y_{min} \leftarrow \infty$ 
5          for each  $S \ni e$ 
6              do  $y' \leftarrow \sum_{e' \in S - \{e\}} y_{e'}$ 
7                  if  $y' < y_{min}$  and  $y' > 0$ 
8                      then  $S_{min} \leftarrow S, y_{min} \leftarrow y'$ 
9           $x_S \leftarrow 1, y_e \leftarrow y_{min}$ 

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This algorithm achieves an approximation factor of f , where f is the maximum number of times a given element is covered. To show this, first observe that it is obvious all elements are eventually covered, and so there are no overpacked sets (more formally, $\forall S : \sum_{e \in S} y_e \leq c(S)$). Thus, the final solution is both primal and dual feasible. Pick approximation factors $\alpha = 1, \beta = f$, thus we have a f -factor approximation.

To show this bound is tight, consider the following example. Let

$$\begin{aligned}
U &= \{e_0, \dots, e_n\} \\
S &= \{\{e_0, e_1\}, \{e_0, e_2\}, \dots, \{e_0, e_{n-1}\}\} \cup U
\end{aligned}$$

$$c(\{e_0, e_i\}) = 1, \forall i \in 1..n-1$$

$$c(U) = 1 + \epsilon, \epsilon > 0$$

The optimal solution is to pick the cover $\{U\}$ with total cost $1 + \epsilon$, while the above algorithm will end up picking all sets in S for a total cost of $n + \epsilon$.