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Assignment 2

CIS 621: Algorithms and Complexity

Problem 1 (5 points) For the linear program below, where $a_1, a_2, a_3, b_1, b_2, b_3, c_1, c_2, d_3, e_1, e_2, e_3$ are non-zero constants, derive (1) its dual linear program and (2) the KKT conditions for the dual linear program.

$$\begin{aligned} \min_{x_1 \geq 0, x_2 \leq 0, x_3} & a_1 x_1 + a_2 x_2 + a_3 x_3 \\ s. \, t. & b_1 x_1 + b_2 x_2 + b_3 x_3 \leq e_1 \,, \\ & c_1 x_1 + c_2 x_2 = e_2 \,, \\ & d_3 x_3 \geq e_3 \,. \end{aligned}$$

Solution Assume α , β , and γ are the dual variables for the three constraints, respectively; also assume δ is the dual variable for $x_1 \geq 0$, and θ is the dual variable for $x_2 \leq 0$. Skipping the intermediate steps of derivation, we show the following dual problem:

$$\max_{\alpha \geq 0, \beta \geq 0, \gamma \geq 0, \delta \geq 0, \theta \geq 0} -e_1 \alpha - e_2 \beta + e_3 \gamma
s.t. a_1 + b_1 \alpha + c_1 \beta - \delta = 0,
a_2 + b_2 \alpha + c_2 \beta + \theta = 0,
a_3 + b_3 \alpha - d_3 \gamma = 0.$$
(1)

Note that this problem can be equivalently rewritten as

For this new version of the problem, let's assume μ , η , ρ are the dual variables for the three constraints, respectively, and also assume π , ϕ , ϵ are the dual variables for $\alpha \geq 0$, $\beta \geq 0$, $\gamma \geq 0$, respectively. Skipping the intermediate steps of derivation, we have the following KKT conditions.

Stationarity:

$$e_1 - b_1 \mu + b_2 \eta + b_3 \rho - \pi = 0$$

$$e_2 - c_1 \mu + c_2 \eta - \phi = 0$$

$$-e_3 - d_3 \rho - \epsilon = 0$$

Complementary slackness:

$$\mu(a_1 + b_1\alpha + c_1\beta) = 0$$

$$\eta(a_2 + b_2\alpha + c_2\beta) = 0$$

$$\pi\alpha = 0$$

$$\phi\beta = 0$$

 $\epsilon \gamma = 0$

Primal feasibility:

$$a_1 + b_1 \alpha + c_1 \beta \ge 0$$

 $a_2 + b_2 \alpha + c_2 \beta \le 0$
 $a_3 + b_3 \alpha - d_3 \gamma = 0$
 $\alpha \ge 0, \beta \ge 0, \gamma \ge 0$

Dual feasibility:

$$\mu \ge 0, \eta \ge 0, \rho \ge 0, \pi \ge 0, \phi \ge 0, \epsilon \ge 0$$

Problem 2 (5 points) For the following problem, where a_{ij} , b_{ij} , c_j , d_i , i = 1, 2, ..., m, j = 1, 2, ..., n are positive constants: (1) show that it is a convex optimization problem; (2) derive its KKT conditions.

min
$$\sum_{i=1}^{m} \sum_{j=1}^{n} a_{ij} x_{ij} + \sum_{i=1}^{m} \sum_{j=1}^{n} b_{ij} ((x_{ij} + 1) \ln(x_{ij} + 1) - x_{ij})$$
s.t.
$$\sum_{i=1}^{m} x_{ij} \ge c_j, j = 1, 2, ..., n,$$

$$\sum_{i=1}^{m} \sum_{j=1}^{n} x_{ij} + d_i \ge \sum_{j=1}^{n} x_{ij} + \sum_{j=1}^{n} c_j, i = 1, 2, ..., m,$$

$$x_{ij} \ge 0, i = 1, 2, ..., m, j = 1, 2, ..., n.$$

Solution (1): Omitted. Use Hessian matrix. All constraints are affine (Slater's condition).

(2): Assume α_j , j=1,2,...,n is the dual variable for the first constraint, β_i , i=1,2,...,m is the dual variable for the second constraint, and γ_{ij} , i=1,2,...,m, j=1,2,...,n is the dual variable for the third constraint. Skipping the intermediate steps of derivation, we have the following KKT conditions.

Stationarity:

$$a_{ij} + b_{ij} \ln(x_{ij} + 1) - \alpha_j - \sum_{i=1}^{m} \beta_i + \beta_i - \gamma_{ij} = 0, i = 1, 2, ..., m, j = 1, 2, ..., n$$

Complementary slackness:

$$\alpha_{j}(c_{j} - \sum_{i=1}^{m} x_{ij}) = 0, j = 1, 2, ..., n$$

$$\beta_{i}(\sum_{j=1}^{n} x_{ij} + \sum_{j=1}^{n} c_{j} - \sum_{i=1}^{m} \sum_{j=1}^{n} x_{ij} - d_{i}) = 0, i = 1, 2, ..., m$$

$$\gamma_{ij}x_{ij} = 0, i = 1, 2, ..., m, j = 1, 2, ..., n$$

Primal feasibility:

$$\sum_{i=1}^{m} x_{ij} \ge c_j, j = 1, 2, ..., n$$

$$\sum_{i=1}^{m} \sum_{j=1}^{n} x_{ij} + d_i \ge \sum_{j=1}^{n} x_{ij} + \sum_{j=1}^{n} c_j, i = 1, 2, ..., m$$

$$x_{ij} \ge 0, i = 1, 2, ..., m, j = 1, 2, ..., n$$

Dual feasibility:

$$\begin{aligned} &\alpha_j \geq 0, j = 1, 2, ..., n \\ &\beta_i \geq 0, i = 1, 2, ..., m \\ &\gamma_{ij} \geq 0, i = 1, 2, ..., m, j = 1, 2, ..., n \end{aligned}$$

Problem 3 (10 points) Given $c, x \in \mathbb{R}^n$, $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, we have Problem P_1

min
$$c^T x$$

s. t. $Ax \leq b$,
 $x_i \in \{0, 1\}, i = 1, 2, ..., n$,

Problem P_2

$$0 \le x_i \le 1, i = 1, 2, ..., n$$

and Problem P_3 (which is an equivalent reformulation of P_1)

min
$$c^T x$$

s. t. $Ax \leq b$,
 $x_i(1-x_i) = 0, i = 1, 2, ..., n$.

Note that the optimal value of P_2 is a lower bound, denoted as L_1 , for the optimal value of P_1 . Now, derive the dual problem, denoted as P_4 , for P_3 . The optimal value of P_4 is also a lower bound, denoted as L_2 , for the optimal value of P_1 . Are L_1 and L_2 equal? Explain why. (Hint: derive the dual problem, denoted as P_5 , for P_2 .)

Solution

(a) The Lagrangian is

$$L(x, \mu, \nu) = c^T x + \mu^T (Ax - b) - \nu^T x + x^T \operatorname{diag}(\nu) x$$
$$= x^T \operatorname{diag}(\nu) x + (c + A^T \mu - \nu)^T x - b^T \mu.$$

Minimizing over x gives the dual function

$$g(\mu,\nu) = \begin{cases} -b^T \mu - (1/4) \sum_{i=1}^n (c_i + a_i^T \mu - \nu_i)^2 / \nu_i & \nu \succeq 0 \\ -\infty & \text{otherwise} \end{cases}$$

where a_i is the *i*th column of A, and we adopt the convention that $a^2/0 = \infty$ if $a \neq 0$, and $a^2/0 = 0$ if a = 0.

The resulting dual problem is

maximize
$$-b^T \mu - (1/4) \sum_{i=1}^n (c_i + a_i^T \mu - \nu_i)^2 / \nu_i$$
 subject to $\nu \succeq 0$.

In order to simplify this dual, we optimize analytically over ν , by noting that

$$\sup_{\nu_{i} \geq 0} \left(-\frac{(c_{i} + a_{i}^{T} \mu - \nu_{i})^{2}}{\nu_{i}} \right) = \begin{cases} (c_{i} + a_{i}^{T} \mu) & c_{i} + a_{i}^{T} \mu \leq 0\\ 0 & c_{i} + a_{i}^{T} \mu \geq 0 \end{cases}$$
$$= \min\{0, (c_{i} + a_{i}^{T} \mu)\}.$$

This allows us to eliminate ν from the dual problem, and simplify it as

maximize
$$-b^T \mu + \sum_{i=1}^n \min\{0, c_i + a_i^T \mu\}$$

subject to $\mu \succeq 0$.

(b) We follow the hint. The Lagrangian and dual function of the LP relaxation re

$$L(x, u, v, w) = c^T x + u^T (Ax - b) - v^T x + w^T (x - \mathbf{1})$$

$$= (c + A^T u - v + w)^T x - b^T u - \mathbf{1}^T w$$

$$g(u, v, w) = \begin{cases} -b^T u - \mathbf{1}^T w & A^T u - v + w + c = 0 \\ -\infty & \text{otherwise.} \end{cases}$$

The dual problem is

maximize
$$-b^T u - \mathbf{1}^T w$$

subject to $A^T u - v + w + c = 0$
 $u \succeq 0, v \succeq 0, w \succeq 0,$

which is equivalent to the Lagrange relaxation problem derived above. We conclude that the two relaxations give the same value.