

ON THE DUAL PAIR OF LP-PROBLEMS IN CANONICAL FORM WITH
 NONNEGATIVE INVERSE MATRIX

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Abstract. The dual pair of LP-problems

$$\min\{z = c^T x \mid Ax \geq b, x \geq 0\}, \quad \max\{w = b^T y \mid A^T y \leq c, y \geq 0\}$$

where A is an $n \times n$ matrix with inverse $A^{-1} \geq 0$, is solvable iff $c^T A^{-1} \geq 0$. Otherwise, the primal objective function z is unbounded on the feasible region. Some special cases of the matrix A are discussed.

1. General conclusions

We consider the dual pair of LP-problems

$$\min\{z = c^T x \mid Ax \geq b, x \geq 0\} \quad (P)$$

$$\max\{w = b^T y \mid A^T y \leq c, y \geq 0\} \quad (D)$$

where A is a given real $n \times n$ matrix with inverse $A^{-1} \geq 0$, $b = [b_i]$ and $c = [c_j]$ are given n -vectors, x and y are the vectors of variables in Euclidean space R^n .

Let N denote the set of integers $\{1, 2, \dots, n\}$, and

$$(A^{-1})_j \text{ denote the } j\text{-th column of } A^{-1}.$$

Define

$$x_0 = \sum_{j \in N} (A^{-1})_j$$

$$y_0 = (A^{-1})^T c$$

$$b_0 = A^{-1} b$$

The assumption $A^{-1} \geq 0$, which means $(A^{-1})_j \geq 0$, $j \in N$ and implies $x_0 \geq 0$, plays an essential role. Almost immediately can be stated that:

- (i) The feasible region T of (P) is unbounded;
- (ii) (P) and (D) are solvable iff $y_0 \geq 0$.

Indeed, we can choose k in N satisfying the condition

$$b_k = \max_{i \in N} \{b_i\}. \quad (1)$$

In the case when $b_k \leq 0$, at least the points of the ray

$$\{x = \lambda x_0 \mid \lambda \geq 0\}$$

satisfy the constraints of (P), and clearly, T is unbounded.

In the opposite case, $b_k > 0$, the points of the half-line

$$\{x = (b_k + \lambda)x_0 \mid \lambda \geq 0\}$$

belong to T , and therefore T must be unbounded.

The proof of (ii) also is evident. If $y_0 \geq 0$, then y_0 is a feasible solution to (D) and, by duality theorem, there exists a pair \hat{x}, \hat{y} of optimal solutions to (P) and (D), respectively. If $y_0 \not\geq 0$, and for example its ℓ -th component is negative, $(y_0)_\ell < 0$, then for the points of the half-line

$$\{x = b_k x_0 + \lambda (A^{-1})_\ell \mid \lambda \geq 0\} \subseteq T$$

we have

$$z = c^T x = b_k c^T x_0 + \lambda (y_0)_\ell \rightarrow -\infty \text{ for } \lambda \rightarrow +\infty.$$

So, in this case (P) has no finite optimal solution and (D) is infeasible.

Now, it is easy to make the following conclusion:

(iii) For (P) and (D) if $y_0 \geq 0$ and $b_0 \geq 0$, then the vectors b_0, y_0 are optimal solutions.

Indeed, b_0 is feasible solution to (P), y_0 is feasible solution to (D) and moreover $z_0 = c^T b_0 = c^T A^{-1} b = ((A^{-1})^T c)^T b = y_0^T b = b^T y_0 = w_0$.

In the case when $y_0 \geq 0$, but $b_0 \not\geq 0$, a pair of optimal solutions \hat{x}, \hat{y} to (P) and (D) can be found applying simplex algorithm for the equivalent standard form of (D):

$$\max\{w = b^T y \mid A^T y + v = c, y \geq 0, v \geq 0\}. \quad (D_g)$$

Since $A^{-1} \geq 0$ is known, it is convenient to start with the basis $B=A^T$ and the corresponding reduced form of (D_s) :

$$\max\{w-b^T y_0 = -b_0^T v \mid y+(A^T)^{-1}v = y_0, \quad y \geq 0, \quad v \geq 0\}.$$

2. Some special cases

a) The trivial case $A=I$. For the identity matrix I the condition $y_0 \geq 0$ means $c \geq 0$, and under this assumption the vectors

$$\hat{x} = [\hat{x}_j]_{nx1}, \quad \text{where } \hat{x}_j = \begin{cases} b_j & \text{if } b_j > 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\hat{y} = [\hat{y}_j]_{nx1}, \quad \text{where } \hat{y}_j = \begin{cases} c_j & \text{if } b_j > 0 \\ 0 & \text{otherwise} \end{cases}$$

are optimal solutions; $\hat{z} = \hat{w} = \sum_{j:b_j > 0} c_j b_j$ is the optimal value of z and w .

b) $A = \frac{1}{n-1}E - I$. The $n \times n$ matrix with elements all equal to 1 is denoted by E . For the inverse of A we have $A^{-1} = E - I \geq 0$ and therefore, the necessary and sufficient conditions for solvability of (P) and (D) can be stated as

$$c^T e \geq c_k \quad (2)$$

where e denotes the vector, whose components are all equal to 1, and k is defined by $c_k = \max_{j \in N} \{c_j\}$.

In the case (2), if $e^T b \geq \max_{i \in N} \{b_i\} = b_l$, then

$$\hat{x} = \beta_0 e - b, \quad \text{where } \beta_0 = e^T b, \quad \text{is an optimal solution to (P),}$$

$$\hat{y} = \gamma_0 e - c, \quad \text{where } \gamma_0 = e^T c, \quad \text{is an optimal solution to (D),}$$

and $\hat{z} = \beta_0 \gamma_0 - c^T b = \hat{w}$ is the common optimal value of z and w . If $\beta_0 < b_l$, then it is useful to continue solving the reduced form of (D):

$$\max\{w - \hat{w} = -(\beta_0 e - b)^T v \mid y + (E - I)v = \hat{y}, \quad y \geq 0, \quad v \geq 0\}.$$

Analogous conclusions can be made for the more general case

$$A = \frac{1}{a}(I - \frac{1}{a+n}E), \quad A^{-1} = aI + E, \quad 0 < |a| \leq 1.$$

Namely,

$$\min\{c_j\} \geq -\frac{1}{a}\gamma_0 \quad \text{if } 0 < a \leq 1$$

or,

$$\max\{c_j\} \leq -\frac{1}{a}\gamma_0 \quad \text{if } -1 \leq a < 0$$

is a necessary and sufficient condition for solvability of (P) and (D) in this case. Moreover, if

$$\min\{b_j\} \geq -\frac{1}{a}\beta_0 \quad \text{for } 0 < a \leq 1,$$

or,

$$\max\{b_j\} \leq -\frac{1}{a}\beta_0 \quad \text{for } -1 \leq a < 0$$

then $\hat{x} = \beta_0 e + ab$, $\hat{y} = \gamma_0 e + ac$ is a pair of optimal solutions.

$$c) \quad A = \begin{bmatrix} -(I+E)_{J_1 J_1} & E_{J_1 J_2} \\ E_{J_2 J_1} & (I-E)_{J_2 J_2} \end{bmatrix}$$

This block form of A is corresponding to the subsets of N defined by:

$$J_1 = \{2j-1, j=1, \dots, m\}$$

$$J_2 = \{2j, j=1, \dots, m\}$$

where

$$n = 2m.$$

For the inverse of A we have

$$A^{-1} = \begin{bmatrix} (E-I)_{J_1 J_1} & E_{J_1 J_2} \\ E_{J_2 J_1} & (E+I)_{J_2 J_2} \end{bmatrix}$$

and so, the condition $c^T A^{-1} \geq 0$ for solvability of (P) and (D) reduces to

$$\gamma_0 \geq \max\{c_{j_1}, -c_{j_2}\}, \quad (3)$$

where

$$\gamma_0 = c_{J_1}^T e_{J_1} + c_{J_2}^T e_{J_2}, \quad c_{j_1} = \max_{j \in J_1} \{c_j\}, \quad c_{j_2} = \min_{j \in J_2} \{c_j\}$$

In the case (3), if

$$\beta_0 \geq \max\{b_{j_1}, -b_{j_2}\},$$

where

$$\beta_0 = b_{J_1}^T e_{J_1} + b_{J_2}^T e_{J_2}, \quad b_{j_1} = \max_{j \in J_1} \{b_j\}, \quad b_{j_2} = \min_{j \in J_2} \{b_j\},$$

then the vectors

$$\begin{bmatrix} \hat{x}_{J_1} \\ \hat{x}_{J_2} \end{bmatrix} = \begin{bmatrix} \beta_0 e_{J_1} - b_{J_1} \\ \beta_0 e_{J_2} + b_{J_2} \end{bmatrix}, \quad \begin{bmatrix} \hat{y}_{J_1} \\ \hat{y}_{J_2} \end{bmatrix} = \begin{bmatrix} \gamma_0 e_{J_1} - c_{J_1} \\ \gamma_0 e_{J_2} + c_{J_2} \end{bmatrix}$$

are optimal solutions, and

$$\hat{z} = \hat{w} = \beta_0 \gamma_0 - c_{J_1}^T b_{J_1} + c_{J_2}^T b_{J_2}$$

is the optimal value of z and w .

If $\beta_0 < \max\{b_{j_1}, -b_{j_2}\}$, then it can be considered and solved the reduces form of (D_s) :

$$\max\{w - \hat{w} = -\hat{x}_{J_1}^T v_{J_1} - \hat{x}_{J_2}^T v_{J_2} \mid \begin{cases} y_{J_1} + (E-I)_{J_1 J_1} v_{J_1} + E_{J_1 J_2} v_{J_2} = \hat{y}_{J_1}, y_{J_1} \geq 0, v_{J_1} \geq 0 \\ y_{J_2} + (E-I)_{J_2 J_1} v_{J_1} + (E+I)_{J_2 J_2} v_{J_2} = \hat{y}_{J_2}, y_{J_2} \geq 0, v_{J_2} \geq 0 \end{cases} \}$$

$$d) \underline{A} = \frac{1}{s} [t_{ij}], \quad \text{where } s = \sum_{j=0}^{n-2} a^j, \quad t_{ij} = \begin{cases} (a^{n-i-s})/a^{i-1}, & i=j \\ a^{n-i-j+1}, & i \neq j \end{cases}$$

for a given real $a \neq 0, -1$. The inverse of A is

$$A^{-1} = E - \text{diag}(a^0, a^1, \dots, a^{n-1});$$

$\text{diag}(a^0, a^1, \dots, a^{n-1})$ denotes the diagonal matrix with diagonal elements a^{i-1} , $i=1, \dots, n$. In the case when

$$0 < |a| < 1,$$

we have $A^{-1} > 0$. Let again $\beta_0 = b^T e$, $\gamma_0 = c^T e$. If $\gamma_0 \geq \max_{j \in N} \{c_j a^{j-1}\}$,

then, (P) and (D) are solvable. Moreover, if $\beta_0 \geq \max_{j \in N} \{b_j a^{j-1}\}$ then $\hat{x} = [\hat{x}_j]_{n \times 1}$, where $\hat{x}_j = \beta_0 + b_j a^{j-1}$, is an optimal solution to (P), $\hat{y} = [\hat{y}_j]_{n \times 1}$, where $\hat{y}_j = \gamma_0 + c_j a^{j-1}$, is an optimal solution to (D), and $\beta_0 \gamma_0 + \sum_{j \in N} b_j c_j a^{j-1}$ is the optimal value of the objective functions.

3. Description of the convex polyhedral cones C_1, C_2, C_3

As we know, the cone $C = \{u \mid u^T A^{-1} \geq 0\}$ for (P) and (D) represents the set of vectors c for which (P) and (D) are solvable. In the case of symmetric matrix A the cone C also contains the vectors b for which $A^{-1}b$ is an optimal solution to (P). Therefore, it is of interest to have an explicit form of C as a sum of its edges,

$$C = \{u = \sum_{j \in J} \mu_j q_j, \mu_j \geq 0, j \in J\}$$

for some subset J of integers.

In the case $C_1 = \{u \mid (E+aI)u \geq 0\}$ $0 < |a| \leq 1$, we get $J=N$, and $q_j, j \in N$, defined as follows:

$$(q_j)_i = \begin{cases} -\frac{1}{a+n-1}, & i \neq j \\ 1, & i = j \end{cases}, \quad j \in N$$

$$\text{In the case } C_2 = \left\{ \begin{bmatrix} u_{J_1} \\ u_{J_2} \end{bmatrix} \mid \begin{bmatrix} (E-I)_{J_1 J_1} & E_{J_1 J_2} \\ E_{J_2 J_1} & (E+I)_{J_2 J_2} \end{bmatrix} \begin{bmatrix} u_{J_1} \\ u_{J_2} \end{bmatrix} \geq \begin{bmatrix} 0_{J_1} \\ 0_{J_2} \end{bmatrix} \right\}$$

where $J_1 = \{2j-1, j=1, \dots, m\}$, $J_2 = \{2j, j=1, \dots, m\}$, $n=2m$, for the vectors $q_j, j \in N = J_1 \cup J_2$ we have

$$(q_{2j-1})_i = \begin{cases} -1, & i \in J_1 - \{2j-1\} \\ -2, & i = 2j-1 \\ 1, & i \in J_2 \end{cases}, \quad (q_{2j})_i = \begin{cases} 1, & i \in J_1 \\ -1, & i \in J_2 - \{2j\} \\ 0, & i = 2j \end{cases}$$

If $n=2m+1$, $J_1=\{2j-1, j=1, \dots, m+1\}$, $J_2=\{2j, j=1, \dots, m\}$, then there are $2m+1$ edges of C_2 corresponding to the vectors

$$q_{ij} = \begin{bmatrix} q_{J_1} \\ q_{J_2} \end{bmatrix}, \quad i \in J_1, \quad j \in J_2$$

where

$$(q_{J_1})_s = \begin{cases} -1, & s=i \\ 0, & s \neq i \end{cases}, \quad (q_{J_2})_s = \begin{cases} 1, & s=j \\ 0, & s \neq j \end{cases}$$

For

$$C_3 = \{u \mid (E - \text{diag}(a^0, a^1, \dots, a^{n-1})) u \geq 0\}$$

we can get the explicit form without additional assumptions in the case of a such that $0 < a < 1$. Then we have

$$(q_j)_i = \begin{cases} -1, & i=j \\ \frac{a^{n-i}}{\sum_{k=0}^{n-2} a^k}, & i \neq j, \quad j \in N. \\ k=0, k \neq n-j \end{cases}$$

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ЗА ДУАЛНИОТ ПАР ЛП-ПРОБЛЕМИ ВО КАНОНИЧНА ФОРМА СО
 НЕНЕГАТИВНА ИНВЕРЗНА МАТРИЦА

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Резиме

Дуалниот пар ЛП-проблеми

$$\min\{z=c^T x \mid Ax \geq b, x \geq 0\}, \max\{w=b^T y \mid A^T y \leq c, y \geq 0\}$$

каде што A е $n \times n$ матрица со инверзна $A^{-1} \geq 0$, е решлив ако и само ако $c^T A^{-1} \geq 0$. Во спротивно, функцијата на целта z на примарната задача е неограничена на допустливата област. Разгледани се неколку посебни случаи на матрица A .