

## A Note on Deterministic Equivalents to Stochastic Linear Programming Problems

JITKA ŽAČKOVÁ

*Summary.* Generalized inverse matrices are used as a tool for a study of two-stage linear program under uncertainty. For a special choice of  $\mathbf{M}$  which represents an extension of the so-called complete problem, a deterministic equivalent is given in the explicit form.

1. We shall deal with the two-stage stochastic linear program (resp. linear program under uncertainty) in its standard form (see e. g. [2, 4])

$$\text{minimize } f(\mathbf{x}) = \mathcal{E}\{\mathbf{c}'\mathbf{x} + \varphi(\mathbf{x}, \mathbf{A}, \mathbf{b})\} \quad (1)$$

subject to

$$\mathbf{A}_1 \mathbf{x} = \mathbf{b}_1, \quad \mathbf{x} \geq 0, \quad (2)$$

where  $\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b})$ , for fixed  $\mathbf{x}, \mathbf{A}, \mathbf{b}$ , denotes the minimal value of

$$\mathbf{q}'\mathbf{y} \quad (3)$$

subject to

$$\mathbf{M}\mathbf{y} = \mathbf{b} - \mathbf{A}\mathbf{x}, \quad \mathbf{y} \geq 0. \quad (4)$$

The elements of matrices  $\mathbf{A}_1(m_1, n)$ ,  $\mathbf{M}(m, p)$ ,  $\mathbf{b}_1(m_1, 1)$ ,  $\mathbf{q}(p, 1)$  are given constants whereas the elements of  $\mathbf{A}(m, n)$  and  $\mathbf{b}(m, 1)$  are random variables with a known joint distribution and elements of  $\mathbf{c}(n, 1)$  are random variables with known finite mean values.

Usually, the problem (1), (2) is investigated under following assumptions:

- (i) the set  $\mathfrak{M} = \{\mathbf{x}: \mathbf{x} \geq 0, \mathbf{A}_1 \mathbf{x} = \mathbf{b}_1\}$  is non-empty and bounded,
- (ii) the set  $\{\mathbf{y}: \mathbf{y} \geq 0, \mathbf{M}\mathbf{y} = \mathbf{A}\mathbf{x} - \mathbf{b}\}$  is non-empty for all (finite) realizations of  $\mathbf{A}, \mathbf{b}$  and for all  $\mathbf{x} \in \mathfrak{M}$ ,
- (iii) the function  $\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b})$  is defined and finite for all realizations of  $\mathbf{A}, \mathbf{b}$  and for all  $\mathbf{x} \in \mathfrak{M}$ .

Using the generalized inverse  $\mathbf{M}^*$  we can write  $\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b})$  (see Charnes, Cooper, Thompson [1], Kall [3]) as

$$\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b}) = \max \mathbf{w}'\mathbf{M}^*(\mathbf{A}\mathbf{x} - \mathbf{b}) + \mathbf{q}'\mathbf{M}^*(\mathbf{b} - \mathbf{A}\mathbf{x}) \quad (5)$$

subject to

$$(\mathbf{E} - \mathbf{M}^*\mathbf{M})\mathbf{w} = (\mathbf{E} - \mathbf{M}^*\mathbf{M})\mathbf{q}, \quad \mathbf{w} \geq 0. \quad (6)$$

The constraints (6) are deterministic and do not involve  $\mathbf{x}$ ; the constrained maximum of  $\mathbf{w}'\mathbf{M}^*(\mathbf{A}\mathbf{x} - \mathbf{b})$  can be found as the unconstrained  $\max_{1 \leq s \leq S} \mathbf{w}'_s \mathbf{M}^*(\mathbf{A}\mathbf{x} - \mathbf{b})$ , where  $\mathbf{w}_s, s = 1, \dots, S$ , are extreme points of the set  $\{\mathbf{w}: \mathbf{w} \geq 0, (\mathbf{E} - \mathbf{M}^*\mathbf{M})\mathbf{w} = (\mathbf{E} - \mathbf{M}^*\mathbf{M})\mathbf{q}\}$ .

Further, using the known property of generalized inverses,  $\mathbf{M}^* \mathbf{M} \mathbf{M}^* = \mathbf{M}^*$ , and expressing any vector  $\mathbf{v}$  as  $\mathbf{v} = \mathbf{M}^* \mathbf{M} \mathbf{v} + (\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{v}$ , the problem becomes:

Find 
$$\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b}) = \max(\mathbf{M}^* \mathbf{M} \mathbf{w})' \mathbf{M}^* (\mathbf{A} \mathbf{x} - \mathbf{b}) + (\mathbf{M}^* \mathbf{M} \mathbf{q})' \mathbf{M}^* (\mathbf{b} - \mathbf{A} \mathbf{x}) \quad (7)$$

subject to

$$\mathbf{M}^* \mathbf{M} \mathbf{w} \geq -(\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q}. \quad (8)$$

We shall give a sufficient condition for assumption (iii) to hold. (For a necessary condition see Kall [2], Theorem 5.)

**Theorem 1.** *Let the set  $\{\mathbf{y}: \mathbf{y} \geq 0, \mathbf{M} \mathbf{y} = \mathbf{z}\}$  be non-empty for all  $\mathbf{z} \in E_m$  and let  $(\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q} \geq 0$ . Then assumption (iii) holds.*

*Proof.* The assumption  $\{\mathbf{y}: \mathbf{y} \geq 0, \mathbf{M} \mathbf{y} = \mathbf{z}\} \neq \emptyset$  for all  $\mathbf{z} \in E_m$  implies the existence of a vector  $\mathbf{t}$  such that  $\mathbf{t} \neq 0, \mathbf{t} \geq 0, \mathbf{M} \mathbf{t} = 0$  (see Kall [2], Theorem 3).

The function  $\mathbf{q}' \mathbf{y}$  attains its minimum subject to (4) if  $\mathbf{q}' \mathbf{t} \geq 0$  holds for all  $\mathbf{t} \neq 0, \mathbf{t} \geq 0$ , satisfying  $\mathbf{M} \mathbf{t} = 0$ . Using the symmetry of  $\mathbf{M}^* \mathbf{M}$ , we get for all  $\mathbf{t}$  possessing the mentioned properties that

$$\mathbf{q}' \mathbf{t} = \mathbf{q}' \mathbf{M}^* \mathbf{M} \mathbf{t} + [(\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q}]' \mathbf{t} = [(\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q}]' \mathbf{t} \geq 0$$

according to the assumption  $(\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q} \geq 0$ .

**Corollary.** *Under the assumptions of Theorem 1, the set*

$$\{\mathbf{w}: \mathbf{M}^* \mathbf{M} \mathbf{w} \geq -(\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q}\}$$

*is non-empty.*

2. For a special choice of  $\mathbf{M}$ , the explicit form  $\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b})$  will be given. That choice of  $\mathbf{M}$  represents an extension of the so-called *complete problem* (see Wets [4]).

Let  $\mathbf{M} = (\mathbf{P}; -\mathbf{P})$ , where  $\mathbf{P}$  is a  $(m, p)$  matrix of rank  $m$ . In this case, the set  $\{\mathbf{y}: \mathbf{M} \mathbf{y} = \mathbf{z}, \mathbf{y} \geq 0\}$  is non-empty for arbitrary  $\mathbf{z} \in E_m$  if and only if the set  $\{\mathbf{u}: \mathbf{P} \mathbf{u} = \mathbf{z}\}$  is non-empty for arbitrary  $\mathbf{z} \in E_m$ ; but the latter assertion is entailed by the assumption  $h(\mathbf{P}) = m$ .

We get

$$\mathbf{M}^* = \mathbf{M}' (\mathbf{M} \mathbf{M}')^{-1} = \frac{1}{2} \begin{pmatrix} \mathbf{P}' (\mathbf{P} \mathbf{P}')^{-1} \\ -\mathbf{P}' (\mathbf{P} \mathbf{P}')^{-1} \end{pmatrix} = \frac{1}{2} \begin{pmatrix} \mathbf{P}^* \\ -\mathbf{P}^* \end{pmatrix},$$

$$\mathbf{M}^* (\mathbf{A} \mathbf{x} - \mathbf{b}) = \frac{1}{2} \begin{pmatrix} \mathbf{P}^* (\mathbf{A} \mathbf{x} - \mathbf{b}) \\ -\mathbf{P}^* (\mathbf{A} \mathbf{x} - \mathbf{b}) \end{pmatrix},$$

$$\mathbf{M}^* \mathbf{M} = \frac{1}{2} \begin{pmatrix} \mathbf{P}^* \mathbf{P} & -\mathbf{P}^* \mathbf{P} \\ -\mathbf{P}^* \mathbf{P} & \mathbf{P}^* \mathbf{P} \end{pmatrix},$$

$$\mathbf{M}^* \mathbf{M} \mathbf{w} = \frac{1}{2} \begin{pmatrix} \mathbf{P}^* \mathbf{P} \mathbf{w}_1 - \mathbf{P}^* \mathbf{P} \mathbf{w}_2 \\ -\mathbf{P}^* \mathbf{P} \mathbf{w}_1 + \mathbf{P}^* \mathbf{P} \mathbf{w}_2 \end{pmatrix}, \quad \text{where } \mathbf{w} = \begin{pmatrix} \mathbf{w}_1 \\ \mathbf{w}_2 \end{pmatrix}, \mathbf{w}_1, \mathbf{w}_2 \in E_p.$$

Denoting  $\frac{1}{2}(\mathbf{P}^* \mathbf{P} \mathbf{w}_1 - \mathbf{P}^* \mathbf{P} \mathbf{w}_2) = \mathbf{w}_0$  we have  $\mathbf{M}^* \mathbf{M} \mathbf{w} = \begin{pmatrix} \mathbf{w}_0 \\ -\mathbf{w}_0 \end{pmatrix}$  and condition (8) becomes

$$-\tilde{\mathbf{q}}_1 \leq \mathbf{w}_0 \leq \tilde{\mathbf{q}}_2 \quad (9)$$

where

$$\tilde{\mathbf{q}} = (\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q} = \begin{pmatrix} \tilde{\mathbf{q}}_1 \\ \tilde{\mathbf{q}}_2 \end{pmatrix}, \quad \tilde{\mathbf{q}}_1, \tilde{\mathbf{q}}_2 \in E_p.$$

The problem is to find

$$\begin{aligned} \varphi(\mathbf{x}, \mathbf{A}, \mathbf{b}) &= \max (\mathbf{M}^* \mathbf{M} \mathbf{w})' \mathbf{M}^* (\mathbf{A} \mathbf{x} - \mathbf{b}) + (\mathbf{M}^* \mathbf{M} \mathbf{q})' \mathbf{M}^* (\mathbf{b} - \mathbf{A} \mathbf{x}) \\ &= \max \mathbf{w}'_0 \mathbf{P}^* (\mathbf{A} \mathbf{x} - \mathbf{b}) + \mathbf{q}'_0 \mathbf{P}^* (\mathbf{b} - \mathbf{A} \mathbf{x}) \end{aligned}$$

subject to (9); here

$$\mathbf{q} = \begin{pmatrix} \mathbf{q}_1 \\ \mathbf{q}_2 \end{pmatrix}, \quad \mathbf{q}_1, \mathbf{q}_2 \in E_p,$$

$$\mathbf{q}_0 = \frac{1}{2} (\mathbf{P}^* \mathbf{P} \mathbf{q}_1 - \mathbf{P}^* \mathbf{P} \mathbf{q}_2) \quad \text{and} \quad \mathbf{M}^* \mathbf{M} \mathbf{q} = \begin{pmatrix} \mathbf{q}_0 \\ -\mathbf{q}_0 \end{pmatrix}.$$

Denote  $\mathbf{P}^*_i$  the  $i$ -th row of  $\mathbf{P}^*$  and  $z_i = \mathbf{P}^*_i (\mathbf{A} \mathbf{x} - \mathbf{b})$ . Now, the announced result is given by

**Theorem 2.** Let  $\mathbf{M} = (\mathbf{P}^* : -\mathbf{P}^*)$ , where  $\mathbf{P}$  is a  $(m, p)$  matrix of rank  $m$ , let  $q_i + q_{p+i} \geq 0$ ,  $i = 1, \dots, p$ . Then

$$\begin{aligned} \varphi(\mathbf{x}, \mathbf{A}, \mathbf{b}) &= \sum_{i=1}^p q_{p+i} [\mathbf{P}^*_i (\mathbf{A} \mathbf{x} - \mathbf{b})]^+ + \sum_{i=1}^p q_i [\mathbf{P}^*_i (\mathbf{A} \mathbf{x} - \mathbf{b})]^- \\ &= \sum_{i=1}^p q_{p+i} z_i^+ + \sum_{i=1}^p q_i z_i^- \end{aligned} \tag{10}$$

which is a convex separable function in  $z_1, \dots, z_p$ .

*Proof.* The condition  $q_i + q_{p+i} \geq 0$ ,  $i = 1, \dots, p$ , (resp.  $\mathbf{q}_1 + \mathbf{q}_2 \geq 0$ ) is necessary and sufficient for the set  $\{\mathbf{w}_0 \in E_p : -\tilde{\mathbf{q}}_1 \leq \mathbf{w}_0 \leq \tilde{\mathbf{q}}_2\}$  be non-empty. The maximum of  $\mathbf{w}'_0 \mathbf{P}^* (\mathbf{A} \mathbf{x} - \mathbf{b})$  is attained when the  $i$ -th component of the vector  $\mathbf{w}_0$  equals to  $\tilde{q}_{p+i}$  for  $\mathbf{P}^*_i (\mathbf{A} \mathbf{x} - \mathbf{b}) \geq 0$  and equals to  $-\tilde{q}_i$  for  $\mathbf{P}^*_i (\mathbf{A} \mathbf{x} - \mathbf{b}) \leq 0$ ,  $i = 1, 2, \dots, p$ , what gives the desired form of  $\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b})$ . The condition  $q_i = q_{p+i} \geq 0$ ,  $i = 1, \dots, p$ , secures the convexity of  $\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b})$  with respect to  $\mathbf{x}$ , too.

Especially, for the complete problem with  $\mathbf{M} = (\mathbf{E}^* : -\mathbf{E})$  ( $\mathbf{E}$  is the identity matrix), we have

$$\mathbf{M}^* = \frac{1}{2} \begin{pmatrix} \mathbf{E} \\ -\mathbf{E} \end{pmatrix}, \quad \mathbf{M}^* \mathbf{M} = \frac{1}{2} \begin{pmatrix} \mathbf{E} & -\mathbf{E} \\ -\mathbf{E} & \mathbf{E} \end{pmatrix}, \quad (\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q} = \begin{pmatrix} \tilde{\mathbf{q}}_1 \\ \tilde{\mathbf{q}}_1 \end{pmatrix}$$

where the  $i$ -th component of the vector  $\tilde{\mathbf{q}}_1$  equals to  $\frac{1}{2}(q_i + q_{m+i})$  and for arbitrary  $\mathbf{v} \in E_{2m}$  we have  $\mathbf{M}^* \mathbf{M} \mathbf{v} = \begin{pmatrix} \mathbf{v}_0 \\ -\mathbf{v}_0 \end{pmatrix}$  where the  $i$ -th component of the vector  $\mathbf{v}_0$  equals to  $\frac{1}{2}(v_i - v_{m+i})$ .

Now, the assumption  $\tilde{\mathbf{q}} = (\mathbf{E} - \mathbf{M}^* \mathbf{M}) \mathbf{q} \geq 0$  of Theorem 1 is precisely the familiar assumption  $q_i + q_{m+i} \geq 0$ ,  $i = 1, \dots, m$ , and it is both necessary and suffi-

cient for (iii). Further,

$$\begin{aligned} \varphi(\mathbf{x}, \mathbf{A}, \mathbf{b}) = \max & \frac{1}{2} \sum_{i=1}^m (w_i - w_{m+i}) \left( \sum_{j=1}^n a_{ij} x_j - b_i \right) \\ & - \frac{1}{2} \sum_{i=1}^m (q_i - q_{m+i}) \left( \sum_{j=1}^n a_{ij} x_j - b_i \right) \end{aligned}$$

subject to

$$-(q_i + q_{m+i}) \leq w_i - w_{m+i} \leq q_i + q_{m+i}, \quad i = 1, \dots, m.$$

Using (10), we get known result

$$\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b}) = \sum_{i=1}^m q_{m+i} \left( \sum_{j=1}^n a_{ij} x_j - b_i \right)^+ + \sum_{i=1}^m q_i \left( \sum_{j=1}^n a_{ij} x_j - b_i \right)^-.$$

The results remain true for matrices  $\mathbf{M}$  obtained from  $(\mathbf{P}; -\mathbf{P})$  by permutation of columns and by multiplication of columns by possibly different scalars. Especially, the following assertion holds:

**Theorem 3.** Let  $\mathbf{M} = (\mathbf{P}; -\mathbf{PD})$  where  $\mathbf{P}$  is a  $(m, p)$  matrix of rank  $m$  and  $\mathbf{D}$  is a diagonal matrix with positive diagonal elements  $d_1, \dots, d_p$ , let

$$q_i + \frac{q_{p+i}}{d_i} \geq 0, \quad i = 1, \dots, p.$$

Then

$$\varphi(\mathbf{x}, \mathbf{A}, \mathbf{b}) = \sum_{i=1}^p (1 + d_i^2) q_{p+i} [\mathbf{R}_i(\mathbf{A}\mathbf{x} - \mathbf{b})]^+ + \sum_{i=1}^p (1 + d_i^2) q_i [\mathbf{R}_i(\mathbf{A}\mathbf{x} - \mathbf{b})]^-,$$

where  $\mathbf{R}_i$ 's are the rows of the matrix  $\mathbf{R} = \mathbf{P}' [\mathbf{P}(\mathbf{E} + \mathbf{D}^2)\mathbf{P}']^{-1}$ .

*Proof.* The proof is similar to that of Theorem 2. Namely

$$\mathbf{M}^* \mathbf{M} = \begin{pmatrix} \mathbf{RP} & -\mathbf{RPD} \\ -\mathbf{DRP} & \mathbf{DRPD} \end{pmatrix},$$

and for an arbitrary vector

$$\mathbf{v} = \begin{pmatrix} \mathbf{v}_1 \\ \mathbf{v}_2 \end{pmatrix}, \quad \mathbf{v}_1, \mathbf{v}_2 \in E_p,$$

we have

$$\mathbf{M}^* \mathbf{M} \mathbf{v} = \begin{pmatrix} \mathbf{v}_0 \\ -\mathbf{D} \mathbf{v}_0 \end{pmatrix} \quad \text{with } \mathbf{v}_0 = \mathbf{R} \mathbf{P} (\mathbf{v}_1 - \mathbf{D} \mathbf{v}_2);$$

the condition (8) becomes

$$\begin{aligned} \mathbf{w}_0 &\geq -\tilde{\mathbf{q}}_1 (= -\mathbf{q}_1 + \mathbf{q}_0) \\ \mathbf{D} \mathbf{w}_0 &\leq \tilde{\mathbf{q}}_2 (= \mathbf{q}_2 + \mathbf{D} \mathbf{q}_0). \end{aligned}$$

### References

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Mrs. Dr. Jitka Žáčková  
Department of Mathematical Statistics  
Charles University  
Sokolovská 83  
Prague 8, Czechoslovakia

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