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On Characterizing weak defining hyperplanes (weak Facets) in DEA with Constant Returns to Scale Technology

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Abstract

The Production Possibility Set (PPS) is defined as a set of inputs and outputs of a system in which inputs can produce outputs. The Production Possibility Set of the Data Envelopment Analysis (DEA) model is contain of two types defining hyperplanes (facets); strong and weak efficient facets. In this paper, the problem of finding weak defining hyperplanes of the PPS of the CCR model is dealt with. However, the equation of strong defining hyperplanes of the PPS of the CCR model can be found in this paper. We state and prove some properties relative to our method. To illustrate the applicability of the proposed model, some numerical examples are finally provided. Our algorithm can easily be implemented using existing packages for operation research, such as GAMS.

Keywords: Data Envelopment Analysis (DEA), Production Possibility Set, Efficient frontier; Weak efficient frontier, Hyperplane.

1 Introduction

The Data Envelopment Analysis (DEA), introduced by Charnes, Cooper and Rhodes (CCR) [6], is a procedure to evaluate the relative efficiency of a set of decision making units (DMU); each uses multiple inputs to produce multiple outputs. The data define a Production Possibility Set (PPS); that can be used to evaluate the efficiency of each of DMUs. The PPS of the DEA models is the smallest set containing the observed DMUs and all feasible input-output level correspondences pertaining to the production process operated by the DMUs. The PPS of the CCR model is the intersection of a finite number of halfspace, whose defining hyperplanes pass through the origin.



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The defining hyperplanes (facets) of the PPS of the CCR model are divided into two categories, including *i*) strong defining hyperplanes, and *ii*) weak defining hyperplanes. One of the problems in DEA is to find the equations of these defining hyperplanes. There are many researches undertaken on the subject of finding *strong* defining hyperplanes (see for example Amirteimoori et al. [1], Amirteimoori et al. [2], M. Davtalab-Olyaie et al. [7], Jahanshahloo et al. [11], Jahanshahloo et al. [16], Wei et al. [15], Hosseinzadeh et al. [9] and Olesen et al. [14]). However, less attention has been paid about finding *weak* defining hyperplanes of the PPS of the CCR model (see Wei et al. [15]). In this paper we provide a method to find *weak* defining hyperplanes of the PPS of the CCR model.

Using the method proposed by Jahanshahloo et al. [11] (with some modifications) we introduce a method to find the equations of *weak* defining hyperplanes of the PPS of the CCR model. The idea to find weak defining hyperplanes is straightforward by adding artificial weak efficient DMUs, named weak efficient virtual DMUs in this paper. The key is how to define these artificial weak efficient DMUs, and it is done by testing all CCR-efficient DMUs by a variance of super-efficiency models (see models (3.5) and (3.6))(after eliminating all CCR-inefficient DMUs from the PPS) and determining all extreme DMUs that lie on the some weak efficient defining hyperplanes. A supporting hyperplane is found to be a weak defining hyperplane if at least one artificial DMU lies on it. Using this method, it is possible to check (i) which CCR-efficient DMUs lie on the extreme rays (edges) of the PPS of the CCR model, (ii) which extreme DMUs lie on the some weak defining hyperplanes of the PPS of the CCR model, (iii) how many weak and strong defining hyperplanes they are on. Also, these hyperplanes are useful in finding the closet target for inefficient DMUs and in sensitivity and stability analysis (see Jahanshahloo et al. [12]). Finally, knowledge of weak facets is required for a thorough analysis, in particular, when the PPS is constructed only of weak facets (see Figure 1 and remark 2.2). These may show the importance of obtaining the weak defining hyperplanes of the PPS of DEA models. Some useful facts related to the properties of models (3.5) and (3.6) are stated and proved. In addition, three numerical examples are provided.

2 Background

Consider a set of n DMUs which is associated with m inputs and s outputs. Particularly, each $DMU_j = (X_j, Y_j)$ $(j \in J = \{1, ..., n\})$ consumes amount $x_{ij}(>0)$ of input i and produces amount $y_{rj}(>0)$ of output r. The production possibility set T, $T \subset \{(X,Y)|X \in E^m, Y \in E^s, X \ge 0, Y \ge 0\}$ is based on postulate sets which are presented with a brief explanation (see Banker [4], Banker et al. [3] and Yu et al. [16]). One of the DEA models to evaluate the relative efficiency of a set of DMUs is the CCR model, which is, proposed by Charnes et al. [6]. The production possibility set (PPS) of the CCR model can be defined as follows:

$$T = \Big\{ (X,Y) | X \geqq \sum_{j \in J} \lambda_j X_j, \ Y \leqq \sum_{j \in J} \lambda_j Y_j, \ \lambda_j \geqslant 0, \ j \in J \Big\}.$$

in which X_i and Y_i are vectors of input and output of DMU_i , respectively.

A face of a polyhedral set is the support set of a supporting hyperplane.

A *facet* of a k-dimensional polyhedral set is a k-1 dimensional face. In fact, any facet of the PPS of the DEA model is a defining hyperplane of the PPS.

Note: A CCR-efficient DMU is said to be extreme DMU; if it lies on the edge of the PPS of the CCR model.

The PPS of the CCR model is depicted in Figure 2. In Figure 1, DMUs D_1 and D_2 are extreme DMUs and CCR-efficient DMU D_3 , that lies on the strong defining hyperplane H_1 is non-extreme DMUs.



The input-oriented CCR model, corresponds to DMU_k , $k \in J$, is given by:

min
$$\theta - \varepsilon (\sum_{i=1}^{m} s_{i}^{-} + \sum_{r=1}^{s} s_{r}^{+})$$

s.t. $\sum_{j \in J} \lambda_{j} y_{rj} - s_{r}^{+} = y_{rk}, \quad r = 1, ..., s$
 $\sum_{j \in J} \lambda_{j} x_{ij} + s_{i}^{-} = \theta x_{ik}, \quad i = 1, ..., m$
 $\lambda_{j} \ge 0, \quad j \in J$
 $s_{i}^{-} \ge 0, \quad i = 1, ..., m$
 $s_{r}^{+} \ge 0, \quad r = 1, ..., s$
 $\theta \quad free$ (2.1)

Also, the output-oriented CCR model, corresponds to DMU_k , $k \in J$, is as follows:

$$\max \quad \varphi + \varepsilon \left(\sum_{i=1}^{m} t_{i}^{-} + \sum_{r=1}^{s} t_{r}^{+} \right)$$

$$s.t. \quad \sum_{j \in J} \lambda_{j} y_{rj} - t_{r}^{+} = \varphi y_{rk}, \quad r = 1, ..., s$$

$$\sum_{j \in J} \lambda_{j} x_{ij} + t_{i}^{-} = x_{ik}, \quad i = 1, ..., m$$

$$\lambda_{j} \ge 0, \quad j \in J$$

$$t_{i}^{-} \ge 0, \quad i = 1, ..., m$$

$$t_{r}^{+} \ge 0, \quad r = 1, ..., s$$

$$\varphi \quad free$$

$$(2.2)$$

where ε is non-Archimedean small and positive number. Models (2.1) and (2.2) are called envelopment forms (with non-Archimedean number).

 DMU_k is said to be strong efficient (CCR-efficient) if and only if for each optimal solutions, either (i) or (ii) happen:

(i)
$$\theta^* = 1$$
 and $(\mathbf{s}^{+*}, \mathbf{s}^{-*}) = (\mathbf{0}, \mathbf{0})$

(ii)
$$\varphi^* = 1$$
 and $(\mathbf{t}^{+*}, \mathbf{t}^{-*}) = (\mathbf{0}, \mathbf{0})$

 DMU_k is said to be weak efficient if and only if for some optimal solutions, either (v) or (iv) happen:

(v)
$$\theta^* = 1$$
 and $(\mathbf{s}^{+*}, \mathbf{s}^{-*}) \neq (\mathbf{0}, \mathbf{0})$

(iv)
$$\varphi^* = 1$$
 and $(\mathbf{t}^{+*}, \mathbf{t}^{-*}) \neq (\mathbf{0}, \mathbf{0})$

Note that if $\theta^* < 1$ and $\phi^* > 1$ then DMU_k is an interior point of the PPS. ¹

Each interior DMU and weak efficient DMU in the CCR model is said to be a CCR-inefficient DMU.

Efficient Frontier is the set of all points (real or virtual DMUs) with efficiency score is equal to unity ($\theta^* = 1$ or $\phi^* = 1$).

Efficient frontier is divided into two categories:

- i) Strong efficient frontier is the set of all (real or virtual) strong efficient (CCR efficient) DMU.
- ii) Weak efficient frontier in which all it's relative interior points (real or virtual DMUs), are weak efficient DMUs.

 $DMU_k = (X_k, Y_k)$ is said to be *non-dominated* if and only if there is not any DMU = (X, Y) (real or virtual) such that: $(-X_k, Y_k) \ge (-X, Y)$ and $(-X_k, Y_k) \ne (-X, Y)$.

We use the following theorem in the next section.



¹(*) is used for optimal solution.

Theorem 2.1. There does not exist any virtual DMU (a member of the PPS) that dominates an DEA-efficient DMU.

Proof. See H. Fukuyama et al. [8]. □

The dual of models (2.1) and (2.2) (without ε i.e. ε =0), which are called multiplier forms, are as models (2.3) and (2.4), respectively:

$$\max \sum_{r=1}^{s} u_{r} y_{rk}$$
s.t.
$$\sum_{r=1}^{s} u_{r} y_{rj} - \sum_{i=1}^{m} v_{i} x_{ij} \leq 0, \quad j = 1, ...n$$

$$\sum_{i=1}^{m} v_{i} x_{ik} = 1,$$

$$u_{r} \geq 0, \qquad r = 1, ..., s$$

$$v_{i} \geq 0, \qquad i = 1, ..., m$$
(2.3)

min
$$\sum_{i=1}^{m} v_{i}x_{ik}$$
s.t.
$$\sum_{i=1}^{m} v_{i}x_{ij} - \sum_{r=1}^{s} u_{r}y_{rj} \ge 0, \quad j = 1, ...n$$

$$\sum_{r=1}^{s} u_{r}y_{rk} = 1,$$

$$u_{r} \ge 0, \qquad r = 1, ..., s$$

$$v_{i} \ge 0, \qquad i = 1, ..., m.$$
(2.4)

 DMU_k is strong efficient if there exists at least one optimal solution (u^*, v^*) for (2.3) with $(u^*, v^*) > \mathbf{0}$, and $u^*y_k = 1$ in which $u^* = (u_1^*, u_2^*, ..., u_s^*)$ and $v^* = (v_1^*, v_2^*, ..., v_m^*)$. Also DMU_k is weak efficient if $u^*y_k = 1$ and no $(u^*, v^*) > \mathbf{0}$ exists. In this case there exist at least one r (or i) so that $u_r^* = 0$ (or $v_i^* = 0$) in all optimal solution of model (2.3) or (2.4) (see example 4.3.). In Figure 2, DMUs D_1 , D_2 and D_3 are strong efficient and D_2' is weak efficient DMU. The evaluation of D_3 and D_2' shows that, model (2.3) has unique optimal solution, which defines two supporting defining hyperplanes H_1 and H_2 passing through D_3 and D_2' , respectively. On the other hand, the evaluation of D_1 indicate that, model (2.3) has alternative optimal solutions, which defines an infinite number of supporting hyperplanes passing through D_1 . Only two of these hyperplanes (i.e. H_1 and H_3) are defining hyperplanes. In fact, if (u^*, v^*) is an unique optimal solution of model (2.3) then $u^{t*}y - v^{t*}x = 0$ is the equation of defining hyperplane of the PPS. In addition, if $(u^*, v^*) > 0$, $u^{t*}y - v^{t*}x = 0$ is the equation of strong defining hyperplane of the PPS (see Definition 2.1). Otherwise, if some components of (u^*, v^*) are zero, then $u^{t*}y - v^{t*}x = 0$ is the equation of weak defining hyperplane of the PPS (see Definition 2.2). A similar discussion holds for model (2.4).

Note that the hyperplanes H_2 and H_3 are weak defining hyperplanes and H_1 is strong defining hyperplane of the PPS of the CCR (see Definitions 2.1 and 2.2.).

In this paper, corresponding to each strong efficient DMU $DMU_j = (x_{1j},...,x_{mj},y_{1j},...,y_{sj})$ we consider virtual DMUs $DMU'_j = (x_{1j},...,x_{lj}+\alpha,...,x_{mj},y_{1j},...,y_{sj})$ and $DMU''_j = (x_{1j},...,x_{mj},y_{1j},...,y_{sj})$

 $y_{1j},...,y_{qj}-\gamma,...,y_{sj}$), in which $\alpha,\gamma>0$. These virtual DMUs are either interior point of the PPS of the CCR model or lie on the some weak defining hyperplanes (see Definition 2.2 and properties 3.2-3.7). In the latter case we call these virtual DMUs as "weak efficient virtual DMU", hereafter. (See DMU D_2 in Figure 2 and DMUs D_1 in Figure 1, for example).

Definition 2.1. The supporting hyperplane $H = \{(x,y) | \bar{u}^t y - \bar{v}^t x = 0, (\bar{u}, \bar{v}) \geq 0, (\bar{u}, \bar{v}) \neq 0\}$ of the PPS of the CCR model is strong defining hyperplane of the PPS if only if it is "defining" and m+s-1 (=the number of outputs and inputs minus one) strong efficient DMUs of the PPS, which are linear independent, lie on H. (In this case, all components of (\bar{u}, \bar{v}) are positive.)



²For definition and properties see Bazaraa et al. [5].

Definition 2.2. The supporting hyperplane $H = \{(x,y) | \bar{u}^t y - \bar{v}^t x = 0, (\bar{u},\bar{v}) \geq 0, (\bar{u},\bar{v}) \neq 0\}$ of the PPS of the CCR model is weak defining hyperplane of the PPS if and only if it is "defining" and m+s-1 weak efficient virtual and strong efficient DMUs of the PPS, which are linear independent, lie on H. (In this case, some components of (\bar{u},\bar{v}) are zero.)

Remark 2.1. In the equation of weak defining hyperplane, if $\bar{u}_q = 0$ (or $\bar{v}_l = 0$), then, this hyperplane is vertical to hyperplane $y_q = 0$ (or $x_l = 0$). In the case of $x_l = 0$, the weak defining hyperplane passes through of l^{th} axis of input.

Remark 2.2. If the number of strong efficient DMUs are less than m+s-1 then all defining hyperplanes of the PPS are weak defining hyperplanes (because, by Definition 2.1, at least m+s-1 strong efficient DMUs are needed to construct strong defining hyperplane).

In this research, we first find the extreme DMUs of the PPS of the CCR model, lying on the some weak defining hyperplanes, and then using models (3.5) and (3.6), the foregone weak efficient virtual DMUs are found. By using them, we find the weak defining hyperplane of the PPS of the CCR model.

Throughout this paper, we must assume that there are not any two strong efficient DMUs as (x,y) and (tx,ty) for all t > 0 and $t \neq 1$. Otherwise, one of them must be deleted.

3 Identifying equations of weak defining hyperplanes

In this section, we identify the equations of weak defining hyperplanes of the PPS of the CCR model in the following way. First, we evaluate each DMU_k , $(k \in J)$ using, models (2.1) or (2.2). Then, we hold all CCR-efficient DMUs, and remove other DMUs. Suppose that the set of all CCR-efficient DMUs is denoted by E. Corresponding to each $DMU_k = (x_{1k}, ..., x_{mk}, y_{1k}, ..., y_{sk}), (k \in E)$, we solve the following models:

min
$$\theta_{l}^{r}$$

s.t. $\sum_{j \in E - \{k\}} \lambda_{j}^{k} x_{lj} \leq \theta_{l}^{k} x_{lk}$

$$\sum_{j \in E - \{k\}} \lambda_{j}^{k} x_{ij} \leq x_{ik}, \quad i = 1, ..., m \quad i \neq l$$

$$\sum_{j \in E - \{k\}} \lambda_{j}^{k} y_{rj} \geq y_{rk}, \quad r = 1, ..., s$$

$$\lambda_{j}^{k} \geq 0, \quad j \in E - \{k\}$$

$$\theta_{l}^{k} \quad free \quad l = 1, ..., m$$
(3.5)

max
$$\varphi_q^k$$

s.t.
$$\sum_{j \in E - \{k\}} \mu_{j}^{k} x_{ij} \leq x_{ik}, \qquad i = 1, ..., m$$

$$\sum_{j \in E - \{k\}} \mu_{j}^{k} y_{qj} \geq \varphi_{q}^{k} y_{qk},$$

$$\sum_{j \in E - \{k\}} \mu_{j}^{k} y_{rj} \geq y_{rk}, \qquad r = 1, ..., s \qquad r \neq q$$

$$\mu_{j}^{k} \geq 0, \qquad \qquad j \in E - \{k\}$$

$$\varphi_{q}^{k} \qquad \qquad free \qquad q = 1, ..., s$$

$$(3.6)$$

The following properties hold for models (3.5) and (3.6). By property 3.1 we can find all extreme CCR-efficient DMUs. The properties 3.3, 3.4, 3.6, and 3.7 provide the necessary and sufficient conditions for lying an extreme CCR-efficient DMU on the weak defining hyperplane.

Property 3.1. In model (3.5) (or (3.6)), if for some l (or q), $\theta_l^{k*} > 1$ (or $\varphi_q^{k*} < 1$) or if for some l(or q), model (3.5) (or model (3.6)) is infeasible, then, DMU_k is an extreme DMU and vice versa.



Proof. Suppose that $\theta_l^{k*} > 1$. First, we show that DMU_k is CCR-efficient. By contradiction let DMU_k is inefficient. At optimality of model (2.1), two cases are happened:

(i)
$$\theta^* = 1$$
 and $(s^{+*}, s^{-*}) \neq 0$

(ii)
$$\theta^* < 1$$

in each cases it can be shown that $\theta_l^{k*} \leq 1$, a contradiction.

Now we show that DMU_k is ,in fact, an extreme CCR-efficient DMU. By contradiction suppose that DMU_k is an non-extreme CCR-efficient. So, the following system has solution:

$$\sum_{j \in E'} \lambda_j x_j = x_k,$$

$$\sum_{j \in E'} \lambda_j y_j = y_k,$$

$$\lambda_j \ge 0, j \in E'$$
(3.7)

Suppose that $(\bar{\lambda}_j, j \in E')$ is a solution of the above system. If $\bar{\lambda}_j = 0$ then, $(\theta_l^k = 1, \lambda_j = \bar{\lambda}_j, j \in E' - \{k\})$ is a solution of model (3.5). Therefor, $\theta_l^{k*} \le 1$, a contradiction. On the other hand if $\bar{\lambda}_j \ne 0$, we rewrite system (3.7) as follows:

$$\sum_{\substack{j \in E' - \{k\} \\ j \in E' - \{k\}}} \bar{\lambda}_j x_j = (1 - \bar{\lambda}_j) x_k,$$

By divided both side of the above equations by $(1 - \bar{\lambda}_j > 0)$; we obtain a solution of model (3.5) as $(\theta_l^k = 1, \lambda_j = \frac{\bar{\lambda}_j}{1 - \bar{\lambda}_j}, j \in E' - \{k\})$. Therefor, $\theta_l^{k*} \leq 1$, a contradiction. Thus, DMU_k is an extreme CCR-efficient DMU. Now, suppose that for some l, model (3.5) is infeasible. In the similar manner, it can be shown that DMU_k is an extreme CCR-efficient DMU. Conversely, suppose that DMU_k is extreme DMU and model (3.5) is feasible. We show that $\theta_l^{k*} \geq 1$. Consider the following corresponding to DMU_k :

$$\min_{s.t.} \quad \frac{\theta_{l}^{'k}}{\sum_{j \in E'}} \lambda_{j}^{k} x_{lj} + s_{l}^{-} = \theta_{l}^{'k} x_{lk}
\sum_{j \in E'} \lambda_{j}^{k} x_{ij} + s_{i}^{-} = x_{ik}, \quad i = 1, ..., m \quad i \neq l
\sum_{j \in E'} \lambda_{j}^{k} y_{rj} - s_{r}^{+} = y_{rk}, \quad r = 1, ..., s
\lambda_{j}^{k} \geq 0, \quad j \in E'
s_{i}^{-} \geq 0, s_{r}^{+} \geq 0 \quad i = 1, ..., m, \quad r = 1, ..., s
\theta_{l}^{'k} \quad free \quad l = 1, ..., m$$
(3.8)

Now suppose that $\theta^*(=1)$, $\theta_l^{'k*}$ and θ_l^{k*} are the optimal objective functions of the models (2.1), (3.8) and (3.5) with respect to DMU_k , respectively. It is not difficult to show that $\theta^* \leq \theta_l^{'k*} \leq \theta_l^{k*}$. Therefor, $\theta_l^{k*} \geq 1$.

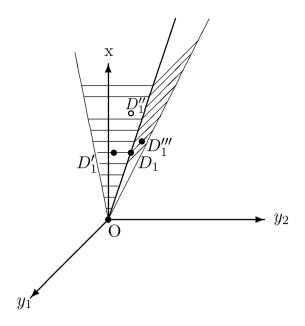


Figure 1: Properties 3.2 and 3.6.

This completes the proof.

Property 3.2. In models (3.5) and (3.6), for each l and q, $\theta_l^{k*} = \varphi_q^{k*} = 1$ if and only if DMU_k is a non-extreme CCR-efficient DMU.

Proof. Omitted.
$$\Box$$

Property 3.3. In a single input case, for each $DMU_k = (x_{1k}, y_{1k}, ..., y_{sk})$, the virtual $DMUDMU'_k = (x_{1k} + \alpha, y_{1k}, ..., y_{sk})$, in which $\alpha > 0$, is an interior point of the PPS of the CCR model.

Proof. First, we add DMU_k' to the PPS and then, evaluate its performance by the input and output-oriented CCR model (see models (2.1) and (2.2)). It is enough to show that $\theta^* < 1$ and $\phi^* > 1$. Consider the input-oriented CCR model corresponding to virtual DMU DMU_k' as follows:

min
$$\theta$$

 $s.t.$
$$\sum_{j \in E} \lambda_{j} x_{1j} + \mu_{k}(x_{1k} + \alpha) \leq \theta(x_{1k} + \alpha),$$

$$\sum_{j \in E} \lambda_{j} y_{rj} + \mu_{k} y_{rk} \geq y_{rk}, \qquad r = 1, ..., s$$

$$\lambda_{j} \geq 0, \qquad j \in E$$

$$\theta \qquad free$$

$$(3.9)$$

 $(\bar{\lambda}_j = 0 (j \neq k), \bar{\lambda}_k = 1, \bar{\mu}_k = 0, \bar{\theta} = \frac{x_{1k}}{x_{1k} + \alpha} (<1))$ is a feasible solution of (3.9). Since model (3.9) has a minimization-type objective function, $\theta^* < 1$; where "*" is used to indicate optimality. In a similar manner, it can be shown that in output-oriented maximization problem, $\phi^* > 1$. Therefore, DMU_k is an interior point of PPS. This completes the proof.

In Figure 1, corresponding to DMU $D_1 = (x_{11}, y_{11}, y_{21})$, virtual DMU $D_1'' = (x_{11} + \alpha, y_{11}, y_{21})$ is an interior point of the PPS.

Property 3.4. In a multiple inputs case, if for some l, model (3.5) is infeasible, then CCR-efficient DMU_k lies on the weak defining hyperplane, which passes through the lth axis of input.



Proof. We show that if for one l, model (3.5) is infeasible, then, virtual DMU

 $DMU'_k = (x_{1k}, ..., x_{(l-1)k}, x_{lk} + \alpha, x_{(l+1)k}, ..., x_{mk}, y_{1k}, ..., y_{sk})$, in which $\alpha > 0$, is on the weak defining hyperplane, which passes through the *l*th axis of input. For this aim, we show that in the performance evaluation of DMU'_k using model (2.2); $\varphi^* = 1$. Consider model (2.2) corresponding to virtual DMU DMU'_k as follows (without ε):

max
$$\varphi$$

s.t.
$$\sum_{j \in E} \lambda_{j} y_{rj} + \mu_{k} y_{rk} \geq \varphi y_{rk}, \qquad r = 1, ..., s$$

$$\sum_{j \in E} \lambda_{j} x_{lj} + \mu_{k} x_{ik} \leq x_{ik}, \qquad i = 1, ..., m, \quad i \neq l$$

$$\sum_{j \in E} \lambda_{j} x_{lj} + \mu_{k} (x_{lk} + \alpha) \leq x_{lk} + \alpha$$

$$\mu_{k}, \lambda_{j} \geq 0, \qquad j \in E$$

$$\varphi \qquad free$$

$$(3.10)$$

By contradiction, suppose that $(\lambda_j^* \ (j \in E), \ \mu_k^*, \ \varphi^*(>1))$ is the optimal solution of (3.10). The constraints of model (3.10) can be written as follows:

$$\sum_{\substack{j \in E - \{k\} \\ \sum_{j \in E - \{k\}}}} \lambda_{j}^{*} y_{rj} > (1 - \lambda_{k}^{*} - \mu_{k}^{*}) y_{rk}, \qquad r = 1, ..., s$$

$$\sum_{\substack{j \in E - \{k\} \\ \sum_{j \in E - \{k\}}}} \lambda_{j}^{*} x_{ij} \le (1 - \lambda_{k}^{*} - \mu_{k}^{*}) x_{ik}, \qquad i = 1, ..., m, \quad i \neq l$$

$$\sum_{\substack{j \in E - \{k\} \\ j \in E - \{k\}}}} \lambda_{j}^{*} x_{lj} \le (1 - \lambda_{k}^{*} - \mu_{k}^{*}) x_{lk} + (1 - \mu_{k}^{*}) \alpha$$
(3.11)

From model (3.11), it is easy to show that $1 - \lambda_k^* - \mu_k^* > 0$. Divide both sides of model (3.11) by $1 - \lambda_k^* - \mu_k^* > 0$ and define $\bar{\mu}_j = \frac{\lambda_j^*}{1 - \lambda_k^* - \mu_k^*}$, $j \in E - \{k\}$; so, model (3.11) becomes as follows:

$$\sum_{\substack{j \in E - \{k\} \\ \sum_{j \in E - \{k\}}} \bar{\mu}_{j} x_{ij} \leq x_{ik}, \qquad i = 1, ..., m, \quad i \neq l \\ \sum_{\substack{j \in E - \{k\} \\ j \in E - \{k\}}} \bar{\mu}_{j} x_{lj} \leq x_{lk} + \beta$$
(3.12)

in which $\beta = \left(\frac{1-\mu_k^*}{1-\lambda_k^*-\mu_k^*}\right)\alpha$. Since $\beta > 0$, there is $\hat{\theta} > 0$ so that $x_{lk} + \beta = \hat{\theta}x_{lk}$; therefore, the constraints of model (3.11) can be rewritten as follows:

$$\sum_{\substack{j \in E - \{k\} \\ \sum_{j \in E - \{k\}}}} \bar{\mu}_j y_{rj} > y_{rk}, \quad r = 1, ..., s$$

$$\sum_{\substack{j \in E - \{k\} \\ j \in E - \{k\}}} \bar{\mu}_j x_{ij} \le x_{ik}, \quad i = 1, ..., m, \quad i \ne l$$

So, $(\bar{\mu}_j \ (j \in E - \{k\}), \hat{\theta})$ is a feasible solution for model (3.5); a contradiction. This implies that $\varphi^* = 1$ i.e. DMU'_k lies on the efficient frontier. Now, since DMU'_k is dominated by CCR-efficient DMU_k , so, DMU'_k lies on the weak efficient frontier (hyperplane). Moreover, it is easy to shows that in the equation of this weak efficient hyperplane; $v_l = 0$ and so by remark 2.1 this hyperplane passes through the *l*th axis of input. This completes the proof.

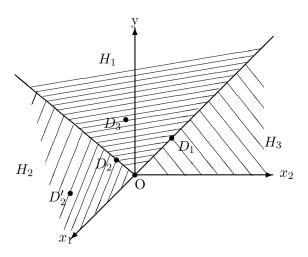


Figure 2: Strong and weak defining hyperplanes of the PPS; Property 3.3

In Figure 2, model (3.5) corresponding to DMU $D_2 = (x_{12}, x_{22}, y_2)$ with l = 1 is infeasible; so, virtual DMU $D'_2 = (x_{12} + \alpha, x_{22}, y_2)$ is on the weak defining hyperplane, which passes through x_1 -axis and vertical to hyperplane x_1 =0.

The following property is, in fact, the converse of property 3.3.

Property 3.5. In a multiple inputs case, if extreme CCR-efficiency DMU DMU_k = $(x_{1k},...,x_{lk},...,x_{nk},y_{1k},...,y_{sk})$ lies on the weak defining hyperplane which passes through the lth axis of input (vertical to hyperplane $x_{l}=0$); then model (3.5) is infeasible.

Proof. By contradiction, suppose that the model (3.5) is feasible. The first constraint of the model (3.5) implies that the optimal solution of the model (3.5) is bounded. Suppose that, $(\theta_l^{k*}, \lambda_j^* \ (j \neq k))$ is an optimal solution of it. Note that the first constraint of the model (3.5) is tight at optimality. We first show that $\theta_l^{k*} > 1$. By contradiction, suppose that $\theta_l^{k*} \le 1$. If $\theta_l^{k*} < 1$ we have:

$$\sum_{j \in E - \{k\}} \lambda_j^{k*} x_{lj} = \theta_l^{k*} x_{lk} < x_{lk}$$

$$\sum_{j \in E - \{k\}} \lambda_j^{k*} x_{ij} \le x_{ik}, \qquad i = 1, ..., m \quad i \ne l$$

$$\sum_{j \in E - \{k\}} \lambda_j^{k*} y_{rj} \ge y_{rk}, \qquad r = 1, ..., s$$
(3.13)

It shows that virtual DMU

$$(\sum_{j \in E - \{k\}} \lambda_j^{k*} x_{lj}, ..., \sum_{j \in E - \{k\}} \lambda_j^{k*} x_{lj}, ..., \sum_{j \in E - \{k\}} \lambda_j^{k*} x_{mj}, \sum_{j \in E - \{k\}} \lambda_j^{k*} y_{1j}, ..., \sum_{j \in E - \{k\}} \lambda_j^{k*} y_{sj})$$

dominates the CCR-efficient DMU_k , a contradiction (see Theorem (3.1)). Now, if $\theta_l^{k*} = 1$, we have:

$$\sum_{j \in E - \{k\}} \lambda_j^{k*} x_{lj} = x_{lk}$$

$$\sum_{j \in E - \{k\}} \lambda_j^{k*} x_{ij} \le x_{ik}, \quad i = 1, ..., m \quad i \ne l$$

$$\sum_{j \in E - \{k\}} \lambda_j^{k*} y_{rj} \ge y_{rk}, \quad r = 1, ..., s$$
(3.14)

At least one of the inequality constraints of (3.14) is a strict inequality, because, otherwise, the CCR-efficient DMU_k , is not extreme DMU. So, $\theta_l^{k*} > 1$. Therefor, there exist $\beta > 0$ so that, $\theta_l^{k*} x_{lk} = x_{lk} + \beta$. This means that, the virtual



DMU $DMU_k' = (x_{1k}, ..., x_{(l-1)k}, x_{lk} + \beta, x_{(l+1)k}, ..., x_{mk}, y_{1k}, ..., y_{sk})$ is, in fact, an observed DMU belongs to the PPS of the CCR model. This is a contradiction. Because, we had been eliminated all the CCR-inefficient DMUs from the PPS of the CCR model. The proof is completed.

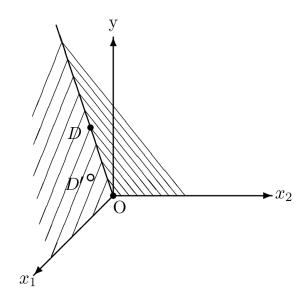


Figure 3: Property 3.5

Property 3.6. In a single output case, for each $DMU_k = (x_{1k},...,x_{mk},y_{1k})$, virtual $DMUDMU'_k = (x_{1k},...,x_{mk},y_{1k}-\gamma)$, in which $\gamma > 0$, is an interior point of the PPS of the CCR model.

Proof. The proof is similar to property 3.2 and so, we omit the details.

In Figure 3, virtual DMU $D' = (x_1, x_2, y_1 - \gamma)$, corresponding to DMU $D = (x_1, x_2, y_1)$ is an interior point of PPS).

Property 3.7. In a multiple outputs case, if for some q, model (3.6) is infeasible, then, CCR-efficient DMU_k lies on the weak defining hyperplane of the PPS, vertical to hyperplane y_q =0.

Proof. The proof is similar to property 3.3 except it can be shown that in the performance evaluation of DMU'_k using model (2.1); $\theta^* = 1$.

In Figure 1, model (3.6) corresponding to DMU $D_1 = (x_{11}, y_{11}, y_{21})$, with q = 2, is infeasible, so, virtual DMU $D'_1 = (x_{11}, y_{11}, y_{21} - \gamma)$ lies on the weak defining hyperplane of the PPS vertical to hyperplane y_2 =0.

The following property is, in fact, the converse of property 3.6.

Property 3.8. In a multiple outputs case, let extreme CCR-efficiency DMU DMU_k = $(x_{1k},...,x_{mk}, y_{1k},...,y_{gk},...,y_{sk})$ lies on the weak defining hyperplane vertical to hyperplane $y_q = 0$; then model (3.6) is infeasible.

Proof. The proof is similar to property 3.4. So, we omit it.

Now, by property 3.1 we can find all extreme DMUs of the PPS of the CCR model and by Properties 3.2, 3.3 and 3.4 we can find all weak efficient virtual DMUs as $DMU_k' = (x_{1k}, ..., x_{(l-1)k}, x_{lk} + \alpha, x_{(l+1)k}, ..., x_{mk}, y_{1k}, ..., y_{sk})$, in which $\alpha > 0$ and by Properties 3.5, 3.6 and 3.7 we can find all weak efficient virtual DMUs as $DMU_k' = (x_{1k}, ..., x_{mk}, y_{1k}, ..., y_{qk} - \beta,, y_{sk})$, in which $\beta > 0$. Put indices of the weak efficient virtual DMUs, in F. Without lose of generality we can



assume that $F \cup E = \{DMU_1, ..., DMU_L\}$. Consider the set $G = \{1, ..., L\}$. Now, we use the method proposed by Jahanshahloo et al. [11] to find all weak defining hyperplanes of the PPS of the CCR model with some modifications. We modify their method for simplifying the process of finding coplanar DMUs. More importantly, we need the following theorems:

Theorem 3.1. Let (x_p, y_p) and (x_q, y_q) be observed DMUs that lie on a strong (or weak) supporting hyperplane, then each convex combination of them is on the same hyperplane.

Proof. See Jahanshahloo et al. [11].

Theorem 3.2. Consider (x_p, y_p) and (x_q, y_q) are two observed DMUs that lie on different hyperplanes (excluding their intersection, if it is not empty). Then every point (virtual DMU) which is obtained by strict convex combination of them is an interior point of PPS. In other words, this virtual DMU is radial inefficient.

Proof. See Jahanshahloo et al. [11]. □

The method is as follows:

Using the above Theorems we first determine all coplanar DMUs in G (i.e. all DMUs that are on the same defining hyperplane). Take a distinct pair DMU_p and DMU_q , where p and q belong to G, and construct a virtual DMU as follows: $DMU_k = \frac{1}{2}DMU_p + \frac{1}{2}DMU_q$. Using the DEA models, we can determine whether or not DMU_k is efficient. In the first case, by Theorem 3.2, DMU_p and DMU_q are on the same hyperplane; in the second case, they are not (by Theorem 3.1). For each $l \in G$, define $G_l = \{j \mid DMU_l \text{ and } DMU_j, j \in G$, are coplanar $\}$. It is obvious that if $\{l_1, l_2, ..., l_p\} \subseteq G_{l_1} \cap G_{l_2} \cap ... \cap G_{l_p}$, then, DMUs $l_1, l_2, ..., l_p$ are coplanar.

Example 3.1. In Figure 4, 1, 2, ..., 6 are six DMUs. Since there exists a plane that binding from 1 and 2, therefore $1 \in G_2$ and $2 \in G_1$ and so on. Then, $G_1 = \{1,2,3\}$, $G_2 = \{1,2,3,4,5\}$, $G_3 = \{1,2,3,4,5\}$, $G_4 = \{2,3,4,5,6\}$, $G_5 = \{2,3,4,5,6\}$, $G_6 = \{4,5,6\}$.

 $\{1,2,3\} \subseteq G_1 \cap G_2 \cap G_3$ therefor, DMUs 1, 2, 3, are coplanar.

 $\{2,3,4,5\} \subseteq G_2 \cap G_3 \cap G_4 \cap G_5$ therefor, DMUs 2, 3, 4, 5 are coplanar.

 $\{4,5,6\} \subseteq G_4 \cap G_5 \cap G_6$ therefor, DMUs 4, 5, 6 are coplanar.

We choose an arbitrary m+s-1 members of G such that $\{l_1,l_2,...,l_{m+s-1}\}\subseteq G_{l_1}\cap G_{l_2}\cap...\cap G_{l_{m+s-1}}$. We call this set $D=\{j_1,\ldots,j_{m+s-1}\}$.

Using D, a hyperplane can be constructed as follows:

$$\begin{vmatrix} x_1 & \cdots & x_m & y_1 & \cdots & y_s \\ x_{1j_1} & \cdots & x_{mj_1} & y_{1j_1} & \cdots & y_{sj_1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ x_{1j_{m+s-1}} & \cdots & x_{mj_{m+s-1}} & y_{1j_{m+s-1}} & \cdots & y_{sj_{m+s-1}} \end{vmatrix} = 0,$$

$$(3.15)$$

where $x_1, \ldots, x_m, y_1, \ldots, y_s$ are variables, $x_{pj_t}, (p = 1, \ldots, m, t = 1, \ldots, m + s - 1)$ is p th input of DMU_{j_t} and y_{qj_t} $(q = 1, \ldots, s; t = 1, \ldots, m + s - 1)$ is q th output of DMU_{j_t} .

Suppose that the equation of the above mentioned hyperplane is in the form of $P^tz = 0$, where $z = (x_1, \dots, x_m, y_1, \dots, y_s)$, and P is the gradient vector of the hyperplane. Considering Theorem 3.3, we can find all equations of weak and strong defining hyperplanes of PPS.

Theorem 3.3. Consider $H = \{z | P^t z = 0\}$ so that $P^t z = 0$ is constructed by (12). Suppose $w = (x_1^w, \dots, x_m^w, y_1^w, \dots, y_s^w)$ is defined as follows:

$$x_i^w = \max\{x_{ij} | j = 1,...,n\} \ i = 1,...,m$$

 $y_r^w = \min\{y_{rj} | j = 1,...,n\} \ r = 1,...,s$



Call w Negative Ideal (NI) if $P^t z_j = 0$ $j \in D$ $P^t z_j \le 0$ $j \in G - D$ $P^t w < 0$

then H is supporting.

Proof. See Jahanshahloo et al. [11].

Now we are in the position to put all together the ingredients of the method.

Summary of finding all Weak Defining Hyperplanes algorithm

- **Step 1**. Evaluate **n** DMUs with a suitable form of models (2.1) and, (2.2). Hold all CCR-efficient DMUs and remove other DMUs. Put indices of this strong efficient DMUs in *E*.
- Step 2. Evaluate each CCR-efficient DMUs with models (3.5) and (3.6). (Note that in the single output case we use model (3.5) and in the single input case we use model (3.6)). Denote the index set of weak efficient virtual DMUs by F. Suppose that $F \cup E = \{DMU_1, ..., DMU_L\}$ and $G = \{1, ..., L\}$.
- Step 3. For each $p, q \in G$ that $p \neq q$, evaluate $DMU_k = \frac{1}{2}DMU_p + \frac{1}{2}DMU_q$ if it is efficient $p \in G_q$ and $q \in G_p$.
- Step 4. For each j (j = 1, ..., L), compute G_j .
- Step 5. Choose arbitrary m+s-1 members of G such that $\{l_1, l_2, ..., l_{m+s-1}\} \subseteq G_{l_1} \cap G_{l_2} \cap ... \cap G_{l_{m+s-1}}$. Call this set as $D = \{j_1, ..., j_{m+s-1}\}$. Construct a hyperplane using equation (3.15). Suppose that the equation of hyperplane is in the form of $P^t z = 0$ where $z = (x_1, ..., x_m, y_1, ..., y_s)$.
- Step 6. If P has any component less than or equal to zero go to step 8, else let $w = (x_1^w, \dots, x_m^w, y_1^w, \dots, y_s^w)$ is defined as follows:

```
\begin{split} x_i^w &= \max\{x_{ij}|\ j=1,\dots,n\}, \quad i=1,\dots,m\\ y_r^w &= \min\{y_{rj}|\ j=1,\dots,n\}, \quad r=1,\dots,s\\ \text{If} \\ P^tz_j &= 0, \quad j\in D\\ P^tz_j &\leq 0, \quad j\in G-D\\ P^tw &< 0,\\ \text{then } P^tz &= 0 \text{ is supporting. Otherwise, go to step 8.} \end{split}
```

- Step 7. If, at least, one of the m+s-1 members of D is a weak efficient virtual DMU, then $P^tz=0$ is weak defining hyperplane. Otherwise, it is strong defining hyperplane.
- Step 8. If another subset of G with m+s-1 members can be found, go to step 5, else stop.



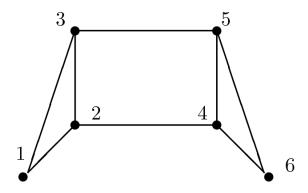


Figure 4: Example 3.1.

4 Numerical Examples

4.1 Single output case

Table 1 shows data for 4 DMUs with two inputs and one output. Using the CCR model (2.1), the CCR-efficient DMUs are determined to be D_1 , D_2 , D_3 . Remove CCR-inefficient DMU D_4 from PPS and solve model (3.5) corresponding to CCR-efficient DMUs D_1 , D_2 and D_3 .

The following results are yielded:

By property 3.1, DMUs D_1 , D_2 and D_3 lie on the extreme rays of the PPS. Model (3.5) corresponding to DMU D_1 with l=2 and DMU D_3 with l=1 is infeasible. So, by property 3.3 weak efficient virtual DMUs $D_5 = D_1' = (1,4+\beta,3)$ and $D_6 = D_3' = (5+\alpha,1,5)$ lie on the weak defining hyperplanes which pass thought the 2th and Ith axis of inputs, respectively. For convenience, we let $\alpha = \beta = 1$. Therefore $E \cup F = \{D_1, D_2, D_3, D_5, D_6\}$ and $G = \{1,2,3,5,6\}$. Note that model (3.5) corresponding to DMU D_2 is feasible. So, by property 3.4, DMU D_2 does not lie on any weak defining hyperplane. Also

$$G_1 = \{1,2,5\}, G_2 = \{2,1,3\}, G_3 = \{3,2,6\}, G_5 = \{1,5\}, G_6 = \{6,3\}.$$

 $\{1,5\} \subseteq G_1 \cap G_5, \{1,2\} \subseteq G_1 \cap G_2,$
 $\{2,3\} \subseteq G_2 \cap G_3, \{3,6\} \subseteq G_3 \cap G_6.$

 H_1 , the first weak hyperplane, is constructed on $D = \{1, 5\}$.

$$\begin{vmatrix} x_1 & x_2 & y \\ 1 & 4 & 3 \\ 1 & 5 & 3 \end{vmatrix} = 0$$
, that yields $y = 3x_1$.

Note that the conditions of Theorem 3.3 are held and H_1 is a weak defining hyperplane including x_2 -axis and vertical to hyperplane $x_2 = 0$ (because the weak efficient virtual DMU, D'_1 , lies on H_1). Here, w = (6,4,2). H_2 , the second weak hyperplane is constructed on $D = \{3,6\}$.

$$\begin{vmatrix} x_1 & x_2 & y \\ 5 & 1 & 5 \\ 6 & 1 & 5 \end{vmatrix} = 0$$
, that yields $y = 5x_2$.

Similarly, the conditions of Theorem 3.3 are held, and hence H_2 is a weak defining hyperplane including x_1 -axis and vertical to hyperplane $x_1 = 0$ (because the weak efficient virtual DMU, D'_3 , lies on H_2).

Using the proposed method, it was found that DMUs D_1 , D_2 and D_3 rest on edges of the PPS. Also, DMUs D_1 and D_3 lie on the edge intersection of strong and weak defining hyperplanes of the PPS. Moreover, using property 3.5 there is no any weak defining hyperplane vertical to hyperplane y = 0.



Table 1: Data of Numerical Example 3.1.

DMU	D1	D2	D3	D4
<i>x</i> ₁	1	2	5	6
x_2	4	2	1	1
y	3	4	5	2

Table 2: Example 2. Multiple input and output.

DMU	D_1	D_2	D_3	D_4	D_5
x_1	2	1	2	4	3
x_2	3	2	2	2	5
y_1	7	3	4	6	5
<i>y</i> ₂	4	5	3	1	2

4.2 Multiple outputs and inputs case

Table 2 shows data for 5 DMUs with two inputs and two outputs. Running model (2.1) (or (2.2)) shows that DMUs D_1 , D_2 , D_4 are CCR-efficient and other DMUs are CCR-inefficient. Applying models (3.5) and (3.6) to each CCR-efficient DMU produces the results reported in Table 3. In Table 3, "INFES" and "FES" denotes "infeasible" and "feasible" respectively. For instance, "INFES" in the first row and the second column means that model (3.5), corresponding to DMU D_1 with l=2, is infeasible. So, by property 3.3, $D_1^2=(2,3+\alpha,7,4)$ is a weak efficient virtual DMU that lies on a weak defining hyperplane passing through x_2 -axis. using property 3.1, it was found that all CCR-efficient DMUs lie on the extreme ray. Using properties 3.3 and 3.6 and the information of Table 3, all weak efficient virtual DMUs can be determined (see Table 4). For simplicity, we choose $\alpha=1$. For simplicity, we rename CCR-efficient and weak efficient virtual DMUs as follows:

$$U_1 = D_1, U_2 = D_2, U_3 = D_4, U_4 = D_1^2, U_5 = D_1^4, U_6 = D_2^1, U_7 = D_2^2, U_8 = D_2^3, U_9 = D_4^1, U_{10} = D_4^4.$$

Therefore, we have:

$$\begin{split} E &= \{U_1, U_2, U_3\}, & F &= \{U_4, U_5, U_6, U_7, U_8, U_9, U_{10}\}, \\ G &= \{U_1, U_2, U_3, U_4, U_5, U_6, U_7, U_8, U_9, U_{10}\}, \\ G_1 &= \{U_1, U_2, U_3, U_4, U_5, U_7, U_{10}\}, & G_2 &= \{U_1, U_2, U_3, U_4, U_6, U_7, U_8, U_9\}, \\ G_3 &= \{U_1, U_2, U_3, U_5, U_6, U_9, U_{10}\}, & G_4 &= \{U_1, U_2, U_4, U_5, U_7\}, \\ G_5 &= \{U_1, U_3, U_4, U_5, U_{10}\}, & G_6 &= \{U_2, U_3, U_6, U_8, U_9\}, \\ G_7 &= \{U_1, U_2, U_4, U_7, U_8\}, & G_8 &= \{U_2, U_6, U_7, U_8\}, \\ G_9 &= \{U_2, U_3, U_6, U_9, U_{10}\}, & G_{10} &= \{U_1, U_3, U_5, U_9, U_{10}\}. \end{split}$$

 H_1 , the first weak hyperplane, is constructed on $D1 = \{1, 2, 4\}$, $D2 = \{1, 2, 7\}$, $D3 = \{2, 1, 7\}$, $D4 = \{2, 4, 7\}$,

$$\begin{vmatrix} x_1 & x_2 & y_1 & y_2 \\ 2 & 3 & 7 & 4 \\ 1 & 2 & 3 & 5 \\ 2 & 4 & 7 & 4 \end{vmatrix} = \begin{vmatrix} x_1 & x_2 & y_1 & y_2 \\ 2 & 3 & 7 & 4 \\ 2 & 4 & 7 & 4 \\ 1 & 3 & 3 & 5 \end{vmatrix} = \begin{vmatrix} x_1 & x_2 & y_1 & y_2 \\ 1 & 2 & 3 & 5 \\ 2 & 3 & 7 & 4 \\ 1 & 3 & 3 & 5 \end{vmatrix} = \begin{vmatrix} x_1 & x_2 & y_1 & y_2 \\ 1 & 2 & 3 & 5 \\ 2 & 4 & 7 & 4 \\ 1 & 3 & 3 & 5 \end{vmatrix} = 0$$

that yields $-23x_1 + 6y_1 + y_2 = 0$.

 H_2 , the second weak hyperplane, is constructed on $D5 = \{1, 10, 3\}$, $D6 = \{1, 3, 5\}$, $D7 = \{1, 5, 10\}$, $D8 = \{3, 5, 10\}$

$$\begin{vmatrix} x_1 & x_2 & y_1 & y_2 \\ 2 & 3 & 7 & 4 \\ 4 & 2 & 6 & .5 \\ 4 & 2 & 6 & 1 \end{vmatrix} = \begin{vmatrix} x_1 & x_2 & y_1 & y_2 \\ 2 & 3 & 7 & 4 \\ 4 & 2 & 6 & 1 \\ 2 & 3 & 7 & 3 \end{vmatrix} = \begin{vmatrix} x_1 & x_2 & y_1 & y_2 \\ 2 & 3 & 7 & 4 \\ 2 & 3 & 7 & 3 \\ 4 & 2 & 6 & .5 \end{vmatrix} = 0$$



that yields $-4x_1 - 16x_2 + 8y_1 = 0$.

The other weak defining hyperplanes are as follows:

```
H_3 is constructed on D9 = \{1,4,5\} that yields -7x_1 + 2y_1 = 0.

H_4 is constructed on D10 = \{2,3,6\}, D11 = \{2,3,9\} and D12 = \{3,6,9\} and D13 = \{2,6,9\}, that yields -27x_2 + 8y_1 + 6y_2 = 0.

H_5 is constructed on D14 = \{3,10,9\} that yields -6x_2 + 2y_1 = 0.

H_6 is constructed on D15 = \{2,6,8\} that yields -5x_2 + 2y_2 = 0.

H_7 is constructed on D14 = \{2,7,8\} that yields -5x_1 - y_2 = 0.
```

One can easily verify that the conditions of Theorem 3.3 are held and H_i , i = 1, ..., 7 are defining. It is worthwhile to note that the aforementioned PPS has only one strong defining hyperplane

$$H_8: -x_1 - 55x_2 + 17y_1 + 12y_2 = 0$$
; which is constructed on $D' = \{1, 2, 3\}$.

The following results can be attained by the aforementioned example:

- There are one strong defining hyperplane and seven weak defining hyperplanes.
- The extreme ray binding from DMU U_1 is the intersection of four defining hyperplanes H_1 , H_2 , H_3 and H_8 .
- The extreme ray binding from DMU U_2 is the intersection of five defining hyperplanes H_1 , H_4 , H_6 , H_7 and H_8 .
- The extreme ray binding from DMU U_3 is the intersection of four defining hyperplanes H_2 , H_4 , H_5 and H_8 .
- All extreme rays are the intersection of strong and weak defining hyperplanes. Also all CCR-efficiency DMUs U_1 , U_2 and U_3 lie on the weak defining hyperplanes.

DMU 2 1 2 D_1 FES INFES FES INFES D_2 INFES **INFES** INFES FES D_4 **INFES** FES FES **INFES**

Table 3: Example 2. The results of evaluation CCR-efficient DMUs by models (3.5) and (3.6).

Table 4: Example 2. Weak efficient virtual DMUs.

DMU	D_1^2	D_1^4	D_2^1	D_2^2	D_{2}^{3}	D_4^1	D_4^4
<i>x</i> ₁	2	2	2	1	1	5	4
x_2	4	3	2	3	2	2	2
V ₁	7	7	3	3	2	6	6
y ₂	4	3	5	5	5	1	0

4.3 Real word data

We evaluated the data of 20 branchs of a bank in Iran using the proposed method. The data was previously analyzed by Amirteimoori et al. [1], (see Table (5)). Running the DEA model (2.1) (or (2.2)) resulted in seven CCR-efficient units as 1, 4, 7, 12, 15, 17 and 20. Using the proposed method, the equations of weak defining hyperplanes were obtained as summarized below:



input output Branch Staff Computer Space m2 Charge Deposits Loans terminals 0.9503 0.70 0.1550 0.1900 0.5214 0.2926 2 0.7962 0.60 1.0000 0.2266 0.6274 0.46243 4 5 0.7982 0.75 0.5125 0.2283 0.9703 0.2606 0.55 1.0000 0.8651 0.2100 0.1927 0.6324 0.2675 0.2333 0.2463 0.8151 0.85 0.7221 6 7 0.5000 0.2069 0.5689 0.8416 0.65 0.6025 0.3500 0.9000 0.7158 0.7189 0.60 0.1824 8 0.7853 0.75 0.1200 0.1250 0.2340 0.2977 0.4756 0.60 0.13500.0801 0.3643 0.2439 0.0486 10 0.6782 0.55 0.5100 0.0818 0.1835 0.4031 1.00 0.3179 0.7112 0.3050 11 0.2117 12 0.8113 0.2550 0.1227 0.9225 0.6279 0.65 13 0.6586 0.3400 0.6452 0.2605 0.85 0.1755 14 0.9763 0.80 0.5400 0.1443 0.5143 0.2433 15 0.6845 0.95 0.45001.0000 0.2617 0.0982 16 0.6127 0.90 0.5250 0.1151 0.4021 0.4641 0.60 0.0900 1.0000 17 1.0000 0.2050 0.1614 0.6337 0.2350 0.0591 0.3492 0.0678 18 0.65 19 0.3715 0.70 0.2375 0.0385 0.1898 0.1112

0.5000

0.1101

0.6145

0.7643

Table 5: Example 3. DMUs' data (extracted from [Amirteimoori et al. [1], p. 689]).

The equations of defining hyperplanes binding from DMU_1 :

0.5827

- 1. $-972780000x_1 35830643110x_3 + 14965298195y_1 + 6971185000y_2 = 0$
- 2. $-8755020000x_1 322475787990x_3 + 134687683755y_1 + 62740665000y_2 = 0$
- 3. $-194556000x_1 7166128622x_3 + 2993059639y_1 + 1394237000y_2 = 0$
- $4. \quad \text{-}1359468485616x_{1} 28494222363367.38x_{3} + 12093263476274y_{1} + 5939111853909y_{2} + 1073596674923.5y_{3} = 0$
- 5. $-97278x_2 2645864x_3 + 1152988y_1 + 497000y_2 = 0$
- 6. 875502 x_2 23812776 x_3 + 10376892 y_1 + 4473000 y_2 =0
- 7. $-194556x_2 52917280x_3 + 23059760y_1 + 9940000y_2 = 0$
- 8. $-1359468485616x_2 22373297128800x_3 + 10143985960000y_1 + 4371989952000y_2 + 726507524000y_3 = 0$
- 9. $-943354x_3 + 388133y_1 + 139000y_2 = 0$

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- 10. $-42450930x_3 + 17465985y_1 + 6255000y_2 = 0$
- $11. \quad \text{-}\ 303325252038x_3 + 125752259800y_1 + 34411902300y_2 + 17703755450y_3 = 0$
- 12. $-286148x_3 + 96226y_1 + 50000y_2 = 0$
- 13. $-143074x_3 + 48113y_1 + 25000y_2 = 0$
- $14. \quad \text{-} \ 100166891292x_3 + 26451830000y_1 + 17352331800y_2 + 49641649000y_3 = 0$

The equations of defining hyperplanes binding from DMU_4 :

- 2. -303325252038x₃+125752259800y₁+3441190230y₂+17703755450y₃=0
- $3. \quad \text{-962972121000} x_1 \text{-328314063906} x_2 + 888323202000 y_1 + 842460036000 y_3 = 0 \\$
- $4. \quad -69807437340x_{1} 109438021302x_{3} + 474497788800y_{1} + 535450401000y_{3} = 0$
- $5. \quad \text{-}30930441000x_1\text{-}48184694626x_2\text{+}29114103500y_2\text{+}34847747500y_3\text{=}0$
- $7. \quad \text{-}6201072000000x_{1}\text{-}63394549535400x_{3} + 18021906285000y_{2} + 7280349255000y_{3} = 0$
- $8. \quad \text{-}3082802028x_2\text{-}1303063920x_3\text{+}2003476500y_2\text{+}702186000y_3\text{=}0$
- 9. $-9415572931881x_2 5266914542760x_3 + 9739254084500y_1 + 5725497127500y_2 + 787058521000y_3 = 0$

- $12. \quad -469929571500x_{1} 5215053344961x_{2} 3555135946650x_{3} + 4096877084250y_{2} + 1430528892750y_{3} = 0\\$
- $14. \quad \text{-}1396148746800x_1 218876042604x_3 + 948995577600y_1 + 1070900802000y_3 = 0 \\$
- $15. \quad -498267154301400x_{1} 675598525370802x_{3} + 412589517921900y_{2} + 312004994380200y_{3} = 0$
- $16. \quad \text{-}7193204732x_2\text{-}3040482480x_3\text{+}4674778500y_2\text{+}1638434000y_3\text{=}0$
- $17. \quad \text{-}203698200x_1\text{-}47097558x_3\text{+}186109800y_3\text{=}0$
- $18. \quad \text{-}209422312020x_{1}\text{-}328314063906x_{3} + 142349336640y_{1} + 160635120300y_{3} = 0\\$
- $19. \quad \text{-}962972121000x_1\text{-}328314063906x_2\text{+}888323202000y_1\text{+}842460036000y_3\text{=}0$
- 20. -2094223120200x₁-328314063906x₂+1423493366400y₁+1606351203000y₂=0
- 21. -588646116x₃+257626800y₁+73971000y₃=0



- 22. $-203698200x_1-47097558x_3+186109800y_3=0$
- 23. $-20670240000x_1-211315165118x_3+60073020950y_2+24267830850y_3=0$
- 24. $-30930441000x_1-48184694626x_2+29114103500y_2+34847747500y_3=0$
- 26. $-49831698600x_1-67566609198x_3+41263078100y_2+31203619800y_3=0$
- $27. \quad -30930441000x_{1} 48184694626x_{2} + 29114103500y_{2} + 34847747500y_{3} = 0$
- 28. $-125676913000x_1-48184694626x_3+49508977400y_2+87532406000y_3=0$
- $29. \quad -125676913000x_{1} 48184694626x_{3} + 49508977400y_{2} + 87532406000y_{3} = 0$
- 30. $-1027600676x_2-434354640x_3+667825500y_2+234062000y_3=0$
- 31. $-44732808x_2 + 20631000y_2 + 11556000y_3 = 0$
- 32. $-203698200x_1-47097558x_3+186109800y_3=0$
- 33. $-203698200x_1-47097558x_3+186109800y_3=0$
- $34. \quad -320990707000x_1 109438021302x_2 + 296107734000y_1 + 280820012000y_3 = 0$
- 35. $-30930441000x_1-48184694626x_2+29114103500y_2+34847747500y_3=0$
- $36. \quad -30930441000x_1 48184694626x_2 + 29114103500y_2 + 34847747500y_3 = 0$
- $38. \quad -429988020954x_2 + 359622357000y_1 + 162565467000y_2 + 64387782000y_3 = 0$
- 39. $-777810000x_1-47097558x_2+931920000y_3=0$
- $40. \quad -30930441000x_1 48184694626x_2 + 29114103500y_2 + 34847747500y_3 = 0$
- $41. \quad \text{-}30930441000x_1\text{-}48184694626x_2\text{+}29114103500y_2\text{+}34847747500y_3\text{=}0$
- 42. $-30930441000x_1 -48184694626x_2 +29114103500y_2 +34847747500y_3 = 0$
- 43. $-30930441000x_1-48184694626x_2+29114103500y_2+34847747500y_3=0$
- $44. \quad \text{-}203698200x_1\text{-}47097558x_3\text{+}186109800y_3\text{=}0$
- $45. \quad -2094223120200x_1 328314063906x_3 + 1423493366400y_1 + 1606351203000y_3 = 0$
- 46. $-196215372x_3+8587560+24657000y_3=0$
- 47. $-62010720000x_1-633945495354x_3+180219062850y_2+72803492550y_3=0$
- $48. \quad -31520343486435x_{1} 192539217627289.8x_{3} + 91136393011015y_{1} + 60913493406765y_{2} + 11617808688185y_{3} = 0$
- 49. $-49831698600x_1-67566609198x_3+41263078100y_2+31203619800y_3=0$
- $50. \quad -125676913000x_1 48184694626x_3 + 49508977400y_2 + 87532406000y_3 = 0$
- $51. \quad \text{-}203698200x_1\text{-}47097558x_3\text{+}186109800y_3\text{=}0$
- 52. $-588646116x_2 + 537594000y_1 + 220161000y_3 = 0$
- $53. \quad \text{-}89793064x_3 + 17110600y_2 + 8035800y_3 = 0$
- $54. \quad -196215372x_2 + 179198000y_1 + 73387000y_3 = 0$
- $55. \quad \text{-}269379192x_3 + 51331800y_2 + 24107400y_3 = 0$
- 56. $-600x_3+126y_3=0$

- $60. \quad \text{-}303325252038x_3 + 125752259800y_1 + 34411902300y_2 + 17703755450y_3 = 0 \\$

The equations of defining hyperplanes binding from DMU_7 :

- 2. $-18000x_1 54330x_2 + 50598y_2 = 0$
- $3. \quad -8890645750000x_1 4649133826135x_2 16924059161800x_3 + 14494910015000y_1 + 13845016272500y_2 = 0$
- 4. $-2163150000x_1 1164847275x_2 1944747000x_3 + 3260731500y_2 = 0$
- 6. $-721050000x_1 388282425x_2 648240000x_3 + 1086910500y_2 = 0$
- 7. 4200x₁ 12677x₂ + 11806y₂=0
- $8. \quad \text{-} \ 9415572931881x_2 \ \text{-} 5266914542760x_3 \ + \ 9739254084500y_1 \ + \ 5725497127500y_2 + \ 787058521000y_3 = 0$
- 9. $3082802028x_2$ $1303063920x_3$ + $2003476500y_2$ + $702186000y_3$ = 0
- $11. \quad \text{-} \ 614548655100x_1 \ \text{-} 818367903423x_3 + 609498131850y_1 + 685617039000y_2 = 0 \\$
- $12. \quad \text{-}\ 477982287300x_1\ \text{-}636508369329x_3\ +\ 474054102550y_1\ +\ 533257697000y_2 = 0$
- 13. $3605250000x_1$ $1941412125x_2$ - $3241245000x_3$ + $5434552500y_2$ =0
- $15. \quad \text{-} \ 614548655100x_1 \ \text{-}818367903423x_3 + 609498131850y_1 + 685617039000y_2 = 0 \\$
- $16. \quad -409699103400x_1 \ -545578602282x_3 \ +406332087900y_1 \ +457078026000y_2 = 0$
- $18. \quad \text{-} \ 1214119364x_2 \ \text{-} 218213520x_3 \ + \ 1074571000y_1 \ + \ 676494000y_2 = 0$
- 19. $9614000x_1$ $5177099x_2$ - $8643320x_3$ + $14492140y_2$ =0
- 21. 8498835548 x_2 -1527494640 x_3 + 7521997000 y_1 + 47354580000 y_2 =0
- 22. 596586144x₂ + 488586000y₁ + 298704000y₂=0

- 23. $-25236750000x_1 13589884875x_2 22688715000x_3 + 38041867500y_2 = 0$
- $24. \quad \text{-} \ 477982287300x_1 \ \text{-} 636508369329x_3 \ + \ 474054102550y_1 \ + \ 533257697000y_2 = 0$
- 25. $-65362500x_1 -46889325x_3 + 70444850y_2=0$
- $26. \quad \text{-} \ 24801090x_1 + 13235861y_2 + 8266595y_3 = 0$
- 27. $-57054x_2 + 33264y_2 + 6000y_3 = 0$
- 28. $1201750000x_1$ $647137375x_2$ - $1080415000x_3$ + $1811517500y_2$ =0
- 29. $-5602500x_1$ $-4019085x_3 + 6038130y_2=0$
- $30. \quad -8412250000x_1 4529961625x_2 7562905000x_3 + 12680622500y_2 = 0$
- 31. $-171162x_2 + 99792y_2 + 18000y_3 = 0$
- $32. \quad -233428299102x_1 + 127007416523y_1 + 93543291008y_2 + 84460030097y_3 = 0$
- 33. $-767039328x_2 + 628182000y_1 + 384048000y_2 = 0$
- 34. $12600x_1$ $38031x_2$ + $3541860y_2$ =0
- 35. $-6000x_1 1811000x_2 + 16866y_2 = 0$
- $36. \quad \text{-}\ 409699103400x_1\ \text{-}545578602282x_3\ +\ 406332087900y_1\ +\ 457078026000y_2 = 0$
- 37. $-5602500x_1$ $-4019085x_3 + 6038130y_2=0$
- $38. \quad \text{-} \ 596586144x_1 + 299539709y_1 + 415833040y_2 = 0 \\$
- 39. $-65362500x_1 -46889325x_3 + 70444850y_2 = 0$
- 40. $-767039328x_1 + 385122483y_1 + 534642480y_2 = 0$
- 41. 10927074276 x_2 -1963921680 x_3 + 9671139000 y_1 + 6088446000 y_2 =0
- 42. $-70980x_2 + 42000y_1 + 38808y_2 = 0$
- 43. $-10140x_2 + 6000y_1 + 5544y_2 = 0$
- 44. $-30930441000x_1 48184694626x_2 + 291141035000y_2 + 34847747500y_3 = 0$
- $45. \quad -156643190500x_1 1738351114987x_2 1185045315550x_3 + 1365625694750y_2 + 476842964250y_3 = 0$
- $46. \quad -49831698600x_1 -67566609198x_3 + 41263078100y_2 + 31203619800y_3 = 0$
- $47. \quad -125676913000x_1 -48184694626x_3 + 49508977400y_2 + 87532406000y_3 = 0$
- $48. \quad \text{-} \ 1027600676x_2 \ \text{-}434354640x_3 + 667825500y_2 + 234062000y_3 = 0$
- $49. \quad \text{-} \ 44732808x_2 + 20631000y_2 + 11556000y_3 = 0$
- 51. $-429988020954x_2 + 359622357000y_1 + 162565467000y_2 + 64387782000y_3 = 0$
- $52. \quad \text{-}\ 7193204732x_2\ \text{-}3040482480x_3\ +\ 4674778500y_2\ +\ 1638434000y_3\ =\ 0$
- $53. \quad -469929571500x_1 5215053344961x_2 3555135946650x_3 + 4096877084250y_2 + 1430528892750y_3 = 0 \\$

The equations of defining hyperplanes binding from DMU_{12} :

- $5. \quad -315203434864350x_1 1925392176272898x_3 + 911363930110150y_1 \\ +609134934067650y_2 \\ +116178086881850y_3 = 0$
- $6. \quad -6201072000000x_1 -63394549535400x_3 +18021906285000y_2 +7280349255000y_3 = 0 \\$

- 9. $-56025000000000x_1-40190850000000x_3+60381300000000y_2=0$
- 10. $-65362500000000x_1-46889325000000x_3+704448500000000y_2 = 0$
- $11. \quad -477982287300000x_1 -636508369329000x_3 +474054102550000y_1 +533257697000000y_2 = 0$

- $16. \quad -21631500000000000x_1 11648472750000000x_2 1944747000000000x_3 + 2607315000000000y_2 = 0\\$

- $19. \quad -12017500000000000x_1 6471373750000000x_2 1080415000000000x_3 + 181151750000000000x_2 = 0 \\$

- 22. 14469168000000000 x₁-1479206155826000000x₃+420511146650000000y₂ +169874815950000000y₃=0

- 25. 1321925000000000x₁-7118511125000000x₂-1188456500000000x₃+19926692500000000y₂=0
 26. 409699103400000x₁ 545578602282000x₂+406332087900000y₁+457078026000000y₂=0
- 27. 39532500000000x₁ 6672000000000x₃ +53210100000000y₂=0
- 28. 291662435700000x₁-752132978400000x₂+431589130800000v₁+407006674500000v₂=0
- 29. $-11859750000000x_1-20016000000000x_3+15963030000000y_2=0$

- 30. $-43749365355000x_1-112819946760000x_3+64738369620000y_1+61051001175000y_2=0$
- 32. $-1976625000000x_1-3336000000000x_3+26605050000000y_2=0$
- $33. \quad \text{-}\ 6588750000000x_1\text{-}11120000000000x_3\text{+}88683500000000y_2\text{=}0$
- 34. $-19766250000000000x_1-333600000000000x_3+26605050000000000y_2=0$

The equations of defining hyperplanes binding from DMU_{15} :

- 1. $-150697636x_1 + 93188442y_1 + 101467310y_3 = 0$
- 2. $-452092908x_1 + 279565326y_1 + 304401930y_3 = 0$
- 3. $-18000x_1 + 12321y_1 = 0$
- 4. $-180x_3+81y_1=0$
- 5. $-291657574659405x_1-752120442850360x_3 + 431581937647820y_1 + 406999891055425y_2 = 0$
- 6. $-43749365355x_1-112819946760x_3+64738369620y_1+61051001175y_2=0$
- 7. $-10927074276x_2-1963921680x_3+9671139000y_1+6088446000y_2=0$
- 8. $-4861040595x_1-12535549640x_3+7193152180y_1+6783444575y_2=0$
- 9. $-2916624357x_1-7521329784x_3+4315891308y_1+4070066745y_2=0$
- $10. \quad -5334387450000x_1 2789480295681x_2 10154435497080x_3 + 8696946009000y_1 + 8307009763500y_2 = 0\\$
- $11. \quad -409699103400x_1 -545578602282x_3 + 406332087900y_1 + 457078026000y_2 = 0$
- $13. \quad \text{-}\ 233428299102x_1 + 127007416523y_1 + 93543291008y_2 + 84460030097y_3 = 0$
- 14. $-767039328x_1 + 385122483y_1 + 534642480y_2 = 0$
- $15. \quad -1214119364x_2 218213520x_3 + 1074571000y_1 + 676494000y_2 = 0$
- $17. \quad \text{-} \ 614548655100x_1 \text{-} 818367903423x_3 + 609498131850y_1 + 685617039000y_2 = 0$
- $18. \quad -8498835548x_2 1527494640x_3 + 7521997000y_1 + 4735458000y_2 = 0$
- 19. 596586144x₁+299539709y₁+415833040y₂ =0
- $20. \quad -62234520250000x_1 -32543936782945x_2 -118468414132600x_3 +101464370105000y_1 +96915113907500y_2 = 0$
- 21. $-477982287300x_1-636508369329x_3+474054102550y_1+533257697000y_2=0$
- 22. $-767039328x_2 + 628182000y_1 + 384048000y_2 = 0$
- 23. $-596586144x_2 + 488586000y_1 + 298704000y_2 = 0$
- $24. \quad -402094816405800x_{1} 774926386660479x_{3} + 506520288276700y_{1} + 385152647434200y_{2} + 169409773083050y_{3} = 0\\$
- $26. \quad -477909287649000x_{1}-338476850586989x_{2}+559061370687500y_{1}+207010083861500y_{2}+360957289402500y_{3}=0\\$
- $28. \quad \text{-} \ 320990707000x_1 \text{-} 109438021302x_2 + 296107734000y_1 + 280820012000y_3 = 0 \\$
- $29. \quad -962972121000x_1 -328314063906x_2 +888323202000y_1 +842460036000y_3 = 0$
- $30. \quad \text{-} \ 698074373400x_1 \text{-} 109438021302x_3 + 474497788800y_1 + 535450401000y_3 = 0 \\$
- 31. $-196215372x_2 + 179198000y_1 + 73387000y_3 = 0$
- $33. \quad \text{-} \ 429988020954x_2 + 359622357000y_1 + 162565467000y_2 + 64387782000y_3 = 0$
- $34. \quad \text{-} \ 588646116x_2 + 537594000y_1 + 220161000y_3 = 0$
- 35. $-196215372x_3 + 85875600y_1 + 24657000y_3 = 0$
- $36. \quad -315203434864350x_{1} 1925392176272898x_{3} + 911363930110150y_{1} + 609134934067650y_{2} + 116178086881850y_{3} = 0$
- $37. \quad -2094223120200x_1 328314063906x_3 + 1423493366400y_1 + 1606351203000y_3 = 0$
- $38. \quad \text{-} \ 588646116x_3 + 257626800y_1 + 73971000 = 0 \\$
- $40. \quad \text{-} \ 194556x_2\text{-}5291728x_3\text{+}2305976y_1\text{+}994000y_2 = 0$
- $41. \quad -97278x_2 2645864x_3 + 1152988y_1 + 497000y_2 = 0$
- $43. \quad \text{-} \ 875502x_2\text{-}23812776x_3\text{+}10376892y_1\text{+}4473000y_2\text{ =}0$
- 44. $-42450930x_3+17465985y_1+6255000y_2=0$
- $45. \quad -1359468485616x_2 -22373297128800x_3 + 10143985960000y_1 + 4371989952000y_2 + 726507524000y_3 = 0$
- $46. \quad \text{-}\ 303325252038x_3 + 125752259800y_1 + 34411902300y_2 + 17703755450y_3 = 0$
- $47. \quad \ -194556000x_1-7166128622x_3+2993059639y_1+1394237000y_2=0$
- 48. $-943354x_3 + 388133y_1 + 139000y_2 = 0$

The equations of defining hyperplanes binding from DMU_{17} :

- $1. \quad -972780000x_1 -35830643110x_3 + 14965298195y_1 + 6971185000y_2 = 0$
- $3. \quad -8755020000x_1 322475787990x_3 + 134687683755y_1 + 62740665000y_2 = 0$
- $5. \quad \text{-} \ 721050000x_1 + 388282425x_2 648249000x_3 + 1086910500y_2 = 0$
- 6. $-658875x_1-1112000x_3+886835y_2=0$

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7. -62234520250000x_1 + 32543936782945x_2 - 118468414132600x_3 + 101464370105000y_1 + 96915113907500y_2 = 0
 8. \quad -62010720000x_1 -633945495354x_3 + 180219062850y_2 + 72803492550y_3 = 0 \\
 9. -6000x_1+18110x_2+16866y_2=0
10. \quad -5334387450000x_1 + 2789480295681x_2 - 10154435497080x_3 + 8696946009000y_1 + 8307009763500y_2 = 0\\
12. -4861040595x_1-12535549640x_3+7193152180y_1+6783444575y_2=0
13. -469929571500x_1 + 5215053344961x_2 - 3555135946650x_2 + 4096877084250y_2 + 1430528892750y_2 = 0
14. \quad \text{-} \ 43749365355x_1\text{-}112819946760x_3\text{+}64738369620y_1\text{+}61051001175y_2\text{=}0
15. -42000x_1+126770x_2+118062y_2=0
16. -395325x_1-667200x_3+532101y_2=0
17. -3605250000x_1+1941412125x_2-3241245000x_3+5434552500y_2=0
18. -315203434864350x_1-1925392176272898x_2+911363930110150y_1+609134934067650y_2+116178086881850y_2=0
19. -2916624357x_1-7521329784x_3+4315891308y_1+4070066745y_2=0
20. \quad -2916580607634645x_{1} - 7521216964053240x_{3} + 4315826569630380y_{1} + 4070005693998825y_{2} = 0
21. \quad -25236750000x_1 + 13589884875x_2 - 22688715000x_3 + 38041867500y_2 = 0
22. \quad -2163150000x_1 + 1164847275x_2 - 1944747000x_3 + 3260731500y_2 = 0
23. -20670240000x_1-211315165118x_3+60073020950y_2+24267830850y_3=0
24. -206681729760000x_1-2112940336014882x_3+600670136479050y_2+242654040669150y_3=0
25. -1976625x_1-3336000x_3+2660505y_2=0
26. \quad \  -194556000x_{1} - 7166128622x_{3} + 2993059639y_{1} + 1394237000y_{2} = 0
27. -18000x_1 + 5433x_2 + 50598y_2 = 0
29. -126000x_1 + 380310x_2 + 354186y_2 = 0
30. -1201750000x_1+647137375x_2-1080415000x_3+1811517500y_2=0
31. -11859750000000x_1-20016000000000x_3+15963030000000y_2=0
32. \quad -10700580000x_{1} - 394137074210x_{3} + 164618280145y_{1} + 76683035000y_{2} = 0
33. \quad -97278x_2 - 2645864x_3 + 1152988y_1 + 497000y_2 = 0
35. -875502x_2-23812776x_3+10376892y_1+4473000y_2=0
36. -8498835548x_2-1527494640x_3+7521997000y_1+4735458000y_2=0
37. -74061281204x_2-13311024720x_3+65548831000y_1+41266134000y_2=0
38. \quad \text{-} \ 70980x_2 + 42000y_1 + 38808y_2 = 0
39. -57054x_2+33264y_2+6000y_3=0
40. \quad -3082802028x_2-1303063920x_3+2003476500y_2+702186000y_3=0
41. -300x_2+180y_2=0
42. -194556x_2-5291728x_3+2305976y_1+994000y_2=0
43. -171162x_2+99792y_2+18000y_3=0
45. -1214119364x_2-218213520x_3+1074571000y_1+676494000y_2=0
46. \quad \  -10927074276x_2-1963921680x_3+9671139000y_1+6088446000y_2=0
47. -10700580x_2-291045040x_3+126828680y_1+54670000y_2=0
48. -1027600676x_2-434354640x_3+667825500y_2+234062000y_3=0
49. -10140x_2+6000y_1+5544y_2=0
50. \quad -89793064x_3\!+\!17110600y_2\!+\!8035800y_3\!=\!0
51. -3000x_3+615y_2=0
52. -2693791920x_3 + 51331800y_2 + 24107400y_3 = 0
53. -143074x_3+48113y_1+25000y_2=0
54. \quad -100166891292x_3 + 26451830000y_1 + 17352331800y_2 + 4964164900y_3 = 0
```

The equations of defining hyperplanes binding from DMU_{20} :

```
1. \quad -1284492135536300x_1 -338476850586989x_3 + 861567778754200y_1 + 351625903900200y_2 + 793901952483300y_3 = 0\\ -1284492135536300x_1 - 338476850586989x_3 + 861567778754200y_1 + 351625903900200y_2 + 793901952483300y_3 = 0\\ -1284492135536300x_1 - 338476850586989x_3 + 861567778754200y_1 + 351625903900200y_2 + 793901952483300y_3 = 0\\ -1284492135536300x_1 - 338476850586989x_3 + 861567778754200y_1 + 351625903900200y_2 + 793901952483300y_3 = 0\\ -1284492135536300x_1 - 338476850586989x_3 + 861567778754200y_1 + 351625903900200y_2 + 793901952483300y_3 = 0\\ -1284492135536300x_1 - 338476850586980x_3 + 861567778754200y_1 + 351625903900200y_2 + 793901952483300y_3 = 0\\ -128449213553600x_1 - 338476850586980x_3 + 861567788754200y_1 + 351625903900200y_2 + 793901952483300y_3 = 0\\ -128449213553600x_1 - 33847685058600x_1 + 361625600x_1 + 36162600x_1 + 3616000x_1 + 3616000x_1 + 3616000x_1 + 3616000x_1 + 3616000x_1 + 3616000x_1 + 36160
   2. -698074373400x_1-109438021302x_3+474497788800y_1+535450401000y_3=0
   3. \quad -125676913000x_1 - 48184694626x_3 + 49508977400y_2 + 87532406000y_3 = 0
   4. \quad \text{-} \ 30930441000x_1\text{-}48184694626x_2\text{+}29114103500y_2\text{+}34847747500y_3\text{=}0
   5. -77781000x_1-47097558x_2+93192000y_3=0
   6. \quad -962972121000x_1 - 328314063906x_2 + 888323202000y_1 + 842460036000y_3 = 0
  7. -2094223120200x_1-328314063906x_3+1423493366400y_1+1606351203000y_3=0
  8. -320990707000x_1-109438021302x_2+296107734000y_1+280820012000y_3=0
   9. \quad \text{-}\ 477909287649000x_{1}\text{-}338476850586989x_{2}\text{+}559061370687500y_{1}\text{+}207010083861500y_{2}\text{+}360957289402500y_{3}\text{=}0
 10. -248010900x_1+132358610y_2+82665950y_3=0
11. \quad -233428299102x_1 + 127007416523y_1 + 93543291008y_2 + 84460030097y_3 = 0
12. -452092908x_1 + 279565326y_1 + 304401930y_3 = 0
13. -150697636x_1 + 93188442y_1 + 101467310y_3 = 0
14. -45858x_1 + 34962y_2 = 0
15. \quad -203698200x_1-47097558x_3+186109800y_3\!=\!0
  16. \quad \text{-} \ 1396148746800x_1\text{-}218876042604x_3\text{+}948995577600y_1\text{+}1070900802000y_3\text{=}0
```

It is worthwhile to note that there are 216 weak defining hyperplanes and there is only one strong defining hyperplane that is constricted on DMUs 4, 7, 12, 15, 17.



5 Conclusions

Until now, less attention has been paid regarding finding *weak* facet of PPS of DEA models (see Wei et al. [15]). Following Jahanshahloo et al. [11], in this paper we proposed a method for finding all weak defining hyperplanes of the PPS of the CCR model. To do this, the performance of each DMUs was firstly evaluated using models (2.1) or (2.2), and all CCR-inefficient cases from the PPS were then removed. By introducing a variance of super-efficient models (see models (3.5) and (3.6)) and using properties 3.2-3.7, the weak efficient virtual DMUs and the strong efficient DMUs are found. A supporting hyperplane was found to be a weak defining hyperplane if at least one weak efficient virtual DMU lies on it. Using the proposed method, one can check which CCR-efficient DMUs lie on the extreme rays (edges) of the PPS of the CCR model; which extreme DMUs lie on the weak defining hyperplanes, and how many defining hyperplanes they are on. In addition, these hyperplanes are useful in sensitivity and stability analysis. Our algorithm can easily be implemented using existing packages for operation research, such as GAMS.

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