This appendix outlines linear programming and its duality relations. Readers are referred to text books such as Gass $(1985)^1$, Charnes and Cooper $(1961)^2$, Mangasarian $(1969)^3$ and Tone $(1978)^4$ for details. More advanced treatments may be found in Dantzig $(1963)^5$, Spivey and Thrall $(1970)^6$ and Nering and Tucker (1993).⁷ Most of the discussions in this appendix are based on Tone (1978).

A.1 LINEAR PROGRAMMING AND OPTIMAL SOLUTIONS

The following problem, which minimizes a linear functional subject to a system of linear equations in nonnegative variables, is called a *linear programming problem*:

$$(P) \quad \min \ z \ = \ cx \tag{A.1}$$

subject to
$$Ax = b$$
 (A.2)

$$x \geq 0,$$
 (A.3)

where $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$ and $c \in \mathbb{R}^n$ are given, and $x \in \mathbb{R}^n$ is the vector of variables to be determined optimally to minimize the scalar z in the objective. c is a row vector and b a column vector. (A.3) is called a nonnegativity constraint. Also, we assume that m < n and rank(A) = m.

A nonnegative vector of variables x that satisfies the constraints of (P) is called *a feasible solution* to the linear programming problem. A feasible solution that minimizes the objective function is called an *optimal solution*.

A.2 BASIS AND BASIC SOLUTIONS

We call a nonsingular submatrix $B \in \mathbb{R}^{m \times m}$ of A a basis of A when it has the following properties: (1) it is of full rank and (2) it spans the space of solutions. We partition A into B and R and write symbolically:

$$A = [B \mid R], \tag{A.4}$$

315

where R is an $(m \times (n - m))$ matrix. The variable vector \boldsymbol{x} is similarly divided into \boldsymbol{x}^B and \boldsymbol{x}^R . \boldsymbol{x}^B is called *basic* and \boldsymbol{x}^R nonbasic. (A.2) can be expressed in terms of this partition as follows:

$$B\boldsymbol{x}^B + R\boldsymbol{x}^R = \boldsymbol{b}. \tag{A.5}$$

By multiplying the above equation by B^{-1} , we have:

$$\boldsymbol{x}^B = B^{-1}\boldsymbol{b} - B^{-1}R\boldsymbol{x}^R. \tag{A.6}$$

Thus, the basic vector x^B is expressed in terms of the nonbasic vector x^R . By substituting this expression into the objective function in (A.1), we have:

$$z = \boldsymbol{c}^{B} B^{-1} \boldsymbol{b} - (\boldsymbol{c}^{B} B^{-1} R - \boldsymbol{c}^{R}) \boldsymbol{x}^{R}.$$
(A.7)

Now, we define a simplex multiplier $\pi \in \mathbb{R}^m$ and simplex criterion $p \in \mathbb{R}^{n-m}$ by

$$\pi = c^B B^{-1} \tag{A.8}$$

$$\boldsymbol{p} = \boldsymbol{\pi} \boldsymbol{R} - \boldsymbol{c}^{\boldsymbol{R}}, \tag{A.9}$$

where π and p are row vectors. The following vectors are called the *basic* solution corresponding to the basis B:

$$\bar{\boldsymbol{x}}^B = B^{-1}\boldsymbol{b} \tag{A.10}$$

$$\bar{\boldsymbol{x}}^R = \boldsymbol{0}. \tag{A.11}$$

Obviously the basic solution is feasible for (A.2) and (A.3).

A.3 OPTIMAL BASIC SOLUTIONS

We call a basis B optimal if it satisfies:

$$\bar{\boldsymbol{x}}^B = B^{-1} \boldsymbol{b} \ge \boldsymbol{0} \tag{A.12}$$

$$\boldsymbol{p} = \boldsymbol{\pi} \boldsymbol{R} - \boldsymbol{c}^{\boldsymbol{R}} \le \boldsymbol{0}. \tag{A.13}$$

Theorem A.1 The basic solution corresponding to an optimal basis is the optimal solution of linear programming (P).

Proof. It is easy to see that $(\bar{x}^B = B^{-1}b, \bar{x}^R = 0)$ is a feasible solution to (P). Furthermore,

$$z = c^B \bar{x}^B - p x^R. \tag{A.14}$$

Hence, by considering $p \leq 0$, we find that z attains its minimum when $x^R = 0$.

The *simplex method* for linear programming starts from a basis, reduces the objective function monotonically by changing bases and finally attains an optimal basis.

A.4 DUAL PROBLEM

Given the linear programming (P) (called the *primal problem*), there corresponds the following *dual problem* with the row vector of variables $y \in \mathbb{R}^m$.

$$(D) \quad \max \ w = \mathbf{yb} \tag{A.15}$$

subject to
$$yA \leq c$$
, (A.16)

and y not otherwise constrained.

Theorem A.2 For each primal feasible solution x and each dual feasible solution y,

$$cx \ge yb.$$
 (A.17)

That is, the objective function value of the dual maximizing problem never exceeds that of the primal minimizing problem.

Proof. By multiplying (A.2) from the left by y, we have

$$yAx = yb. \tag{A.18}$$

By multiplying (A.16) from the right by x and noting $x \ge 0$, we have:

$$yAx \le cx. \tag{A.19}$$

Comparing (A.18) and (A.19),

$$cx \ge yAx = yb. \tag{A.20}$$

Corollary A.1 If a primal feasible x^0 and a dual feasible y^0 satisfy

$$cx^0 = y^0 b, \tag{A.21}$$

then x^0 is optimal for the primal and y^0 is optimal for its dual.

Theorem A.3 (Duality Theorem) (i) In a primal-dual pair of linear programs, if either the primal or the dual problem has an optimal solution, then the other does also, and the two optimal objective values are equal.

(ii) If either the primal or the dual problem has an unbounded solution, then the other has no feasible solution. (iii) If either problem has no solution then the other problem either has no solution or its solution is unbounded.

Proof. (i) Suppose that the primal problem has an optimal solution. Then there exists an optimal basis B and $p = \pi R - c^R \leq 0$ as in (A.13). Thus,

$$\pi R \le c^R. \tag{A.22}$$

However, multiplying (A.8) on the right by B,

$$\pi B = c^B. \tag{A.23}$$

Hence,

$$\pi A = \pi \left[B|R \right] \le \left[c^B | c^R \right] = c. \tag{A.24}$$

Consequently,

$$\pi A \le c. \tag{A.25}$$

This shows that the simplex multiplier π for an optimal basis to the primal is feasible for the dual problem. Furthermore, it can be shown that π is optimal to the dual problem as follows: The basic solution $(\bar{x}^B = B^{-1}b, \bar{x}^R = 0)$ for the primal basis B has the objective value $z = c^B B^{-1}b = \pi b$, while π has the dual objective value $w = \pi b$. Hence, by Corollary A.1, π is optimal for the dual problem. Conversely, it can be demonstrated that if the dual problem has an optimal solution, then the primal problem does also and the two objective values are equal, by transforming the dual to the primal form and by observing its dual. (See Gass, *Linear Programming*, pp. 158-162, for details).

(ii) (a) If the objective function value of the primal problem is unbounded below and the dual problem has a feasible solution, then by Theorem A.2,

$$w = yb \le -\infty. \tag{A.26}$$

Thus, we have a contradiction. Hence, the dual has no feasible solution.

(b) On the other hand, if the objective function value of the dual problem is unbounded upward, it can be shown by similar reasoning that the primal problem is not feasible.

(iii) To demonstrate (iii), it is sufficient to show the following as an example in which both primal and dual problems have no solution.

 $\begin{array}{c|cccc} <\!\! \operatorname{Primal>} & \min & -x & <\!\! \operatorname{Dual>} & \max & y \\ & \operatorname{subject to} & 0 \times x = 1 & & \operatorname{subject to} & y \times 0 \leq -1 \\ & & x \geq 0 \end{array}$

where x and y are scalar variables.

A.5 SYMMETRIC DUAL PROBLEMS

The following two LPs, (P1) and (D1), are mutually dual.

$$\begin{array}{rcl} (P1) & \min & z &= & cx \\ \text{subject to} & Ax &\geq & b \\ & x &\geq & 0. \end{array}$$
 (A.27)

$$\begin{array}{rcl} (D1) & \max & w &= yb\\ \text{subject to} & yA &\leq c & (A.28)\\ & y &> 0. & (A.29) \end{array}$$

The reason is that, by introducing a nonnegative slack $\lambda \in \mathbb{R}^m$, (P1) can be rewritten as (P1') below and its dual turns out to be equivalent to (D1).

$$(P1') \quad \min \ z = cx$$

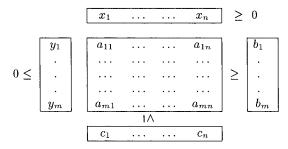
subject to $Ax - \lambda = b$
 $x \ge 0, \ \lambda \ge 0.$ (A.30)

This form of mutually dual problems can be depicted as Table A.1, which is expressed verbally as follows:

For the inequality $\geq (\leq)$ constraints of the primal (dual) problem, the corresponding dual (primal) variables must be nonnegative. The constraints of the dual (primal) problem are bound to inequality $\leq (\geq)$. The objective function is to be maximized (minimized).

This pair of LPs are called symmetric primal-dual problems. The duality theorem above holds for this pair, too.





A.6 COMPLEMENTARITY THEOREM

Let us transform the symmetric primal-dual problems into equality constraints by introducing nonnegative slack variables $\lambda \in \mathbb{R}^m$ and $\mu \in \mathbb{R}^n$, respectively.

$$(P1') \quad \min \ z = cx$$

subject to $Ax - \lambda = b$
 $x \ge 0, \ \lambda \ge 0.$
$$(D1') \quad \max \ w = yb$$

subject to $yA + \mu = c$
 $y \ge 0, \ \mu \ge 0.$ (A.31)
(A.32)

Then, the optimality condition in Duality Theorem A.3 can be stated as follows:

Theorem A.4 (Complementarity Theorem) Let (x, λ) and (y, μ) be feasible to (P1') and (D1'), respectively. Then, (x, λ) and (y, μ) are optimal

to (P1') and (D1') if and only if it holds:

$$\boldsymbol{\mu}\boldsymbol{x} = \boldsymbol{y}\boldsymbol{\lambda} = \boldsymbol{0}.\tag{A.33}$$

Proof.

From $\mu x = 0$, we have $(c - yA)x = 0 \Rightarrow cx = yAx$. (A.34)

From
$$y\lambda = 0$$
, we have $y(Ax - b) = 0 \Rightarrow yAx = yb$. (A.35)

Thus, cx = yb. By the duality theorem, x and y are optimal for the primal and the dual, respectively.

By (A.33), we have

$$\sum_{j=1}^{n} \mu_j x_j = 0, \quad \sum_{i=1}^{m} y_i \lambda_i = 0.$$
 (A.36)

By nonnegativity of each term in these two expressions,

$$\mu_j x_j = 0 \quad (j = 1, \dots, n)$$
 (A.37)

$$y_i \lambda_i = 0. \ (i = 1, \dots, m)$$
 (A.38)

Thus, either the variable μ_j or the variable x_j must be zero for each j and either y_i or λ_i must be zero for each i. We called this property *complementarity*.

A.7 FARKAS' LEMMA AND THEOREM OF THE ALTERNATIVE

Theorem A.5 (Farkas' Lemma, Theorem of the Alternative) For each $(m \times n)$ matrix A and each vector $\mathbf{b} \in \mathbb{R}^m$, either

(I)
$$Ax = b$$
 $x \ge 0$

has a solution $x \in \mathbb{R}^n$ or

$$(II) \quad \boldsymbol{y}A \leq \boldsymbol{0} \quad \boldsymbol{y}\boldsymbol{b} > 0$$

has a solution $y \in \mathbb{R}^m$ but never both.

Proof. For (I), we consider the following the primal-dual pair of LPs:

(P2) min
$$z = 0x$$

 $Ax = b$
 $x \ge 0$
(D2) max $w = yb$
 $yA \le 0$.

If (I) has a feasible solution, then it is optimal for (P2) and hence, by the duality theorem, the optimal objective value of (D2) is 0. Therefore, (II) has no solution.

On the other hand, (D2) has a feasible solution y = 0 and is not infeasible. Hence, if (P2) is infeasible, (D2) is unbounded upward. Thus, (II) has a solution.

A.8 STRONG THEOREM OF COMPLEMENTARITY

Theorem A.6 For each skew-symmetric matrix $K (= -K^T) \in \mathbb{R}^{n \times n}$, the inequality

$$Kx \ge 0, \quad x \ge 0 \tag{A.39}$$

has a solution \bar{x} such that

$$K\bar{\boldsymbol{x}} + \bar{\boldsymbol{x}} > \boldsymbol{0}. \tag{A.40}$$

Proof. Let e_j (j = 1, ..., n) be the *j*-th unit vector and the system (P_j) be,

$$(P_j) Kx \ge 0 x \ge 0, e_j x > 0.$$

If (P_i) has a solution $\bar{x}^j \in \mathbb{R}^n$, then we have:

 $(K\bar{x}^{j})_{j} \ge 0, \ \bar{x}^{j} \ge 0, \ e_{j}\bar{x}^{j} = (\bar{x}^{j})_{j} > 0$

and hence

$$(K\bar{x}^j)_j + (\bar{x}^j)_j > 0.$$

If (P_j) has no solution, then by Farkas' lemma the following system has a solution $\bar{v}^j \in \mathbb{R}^n$, $\bar{w}^j \in \mathbb{R}^n$.

$$(D_j)$$
 $K oldsymbol{v} = oldsymbol{e}_j + oldsymbol{w}$
 $oldsymbol{v} \ge oldsymbol{0}, \quad oldsymbol{w} \ge oldsymbol{0}.$

This solution satisfies:

$$(K\bar{v}^j)_j = 1 + (\bar{w}^j)_j > 0$$

and hence

$$(K\bar{\boldsymbol{v}}^j)_j + (\bar{\boldsymbol{v}}^j)_j > 0.$$

Since, for each j = 1, ..., n, either \bar{x}^j or \bar{v}^j exists, we can define a vector \bar{x} by summing over j. Then \bar{x} satisfies:

$$K\bar{x} + \bar{x} > 0.$$

Let a primal-dual pair of LPs with the coefficient $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$ and $c \in \mathbb{R}^n$ be (P1') and (D1') in Section A.6. Suppose they have optimal solutions $(\bar{x}, \bar{\lambda})$ for the primal and $(\bar{y}, \bar{\mu})$ for the dual, respectively. Then, by the complementarity condition in Theorem A.4, we have:

$$\bar{\mu}_j \bar{x}_j = 0 \ (j = 1, \dots, n)$$
 (A.41)

$$\bar{y}_i \bar{\lambda}_i = 0 \quad (i = 1, \dots, m) \tag{A.42}$$

However, a stronger theorem holds:

Theorem A.7 (Strong Theorem of Complementarity) The primal-dual pair of LPs (P1') and (D1') have optimal solutions such that, in the complementarity condition (A.41) and (A.42), if one member of the pair is 0, then the other is positive.

Proof. Observe the system:

We define a matrix K and a vector \boldsymbol{w} by:

$$K = \begin{pmatrix} O & A & -\mathbf{b} \\ -A^T & O & \mathbf{c}^T \\ \mathbf{b}^T & -\mathbf{c} & 0 \end{pmatrix}$$
$$\mathbf{w}^T = \left(\mathbf{y}^T, \ \mathbf{x}, \ r\right)^T.$$

Then, by Theorem A.6, the system

$$Kw \ge 0, w \ge 0$$

has a solution $\widetilde{\boldsymbol{w}}^T = \left(\widetilde{\boldsymbol{y}}^T, \ \widetilde{\boldsymbol{x}}, \ \widetilde{\boldsymbol{r}}\right)^T$ such that

 $K\widetilde{w} + \widetilde{w} > 0.$

This results in the following inequalities:

$$A\widetilde{\boldsymbol{x}} - \widetilde{\boldsymbol{r}}\boldsymbol{b} + \widetilde{\boldsymbol{y}}^T > \boldsymbol{0} \tag{A.43}$$

$$-\widetilde{y}A + \widetilde{r}c + \widetilde{x}^T > \mathbf{0} \tag{A.44}$$

$$\widetilde{y}b - c\widetilde{x} + \widetilde{r} > 0 \tag{A.45}$$

We have two cases for \tilde{r} .

(i) If $\tilde{r} > 0$, we define \bar{x} and $\bar{\lambda}$ by

$$\bar{x} = \tilde{x}/\tilde{r}, \ \bar{y} = \tilde{y}/\tilde{r}$$
 (A.46)

$$\bar{\lambda} = A\bar{x} - b, \quad \bar{\mu} = c - \bar{y}A.$$
 (A.47)

Then, $(\bar{x}, \bar{\lambda})$ is a feasible solution of (P1') and $(\bar{y}, \bar{\mu})$ is a feasible solution of (D1'). Furthermore, $\bar{y}b \geq c\bar{x}$. Hence, these solutions are optimal for the primal-dual pair LPs. In this case, (A.43) and (A.44) result in

$$\dot{\lambda} + \bar{y} > 0 \tag{A.48}$$

$$\bar{\boldsymbol{\mu}} + \bar{\boldsymbol{x}} > \boldsymbol{0}. \tag{A.49}$$

Thus, strong complementarity holds as asserted in the theorem. (*ii*) If $\tilde{r} = 0$, it cannot occur that both (P1') and (D1') have feasible solutions. The reason is: if they have feasible solutions x^* and y^* , then

$$Ax^* \ge b, x^* \ge 0, y^*A \le c, y^* \ge 0.$$
 (A.50)

Hence, we have:

$$c\widetilde{x} \ge y^* A \widetilde{x} \ge 0 \ge \widetilde{y} A x^* \ge \widetilde{y} b.$$
(A.51)

This contradicts (A.45) in the case $\tilde{r} = 0$. Thus, the case $\tilde{r} = 0$ cannot occur.

A.9 LINEAR PROGRAMMING AND DUALITY IN GENERAL FORM

As a more general LP, we consider the case when there are both nonnegative variables $x^1 \in \mathbb{R}^k$ and sign-free variables $x^2 \in \mathbb{R}^{n-k}$ and both inequality and equality constraints are to be satisfied as follows:

(LP) min
$$z = c^{1}x^{1} + c^{2}x^{2}$$

subject to $A_{11}x^{1} + A_{12}x^{2} \ge b^{1}$ (A.52)
 $A_{21}x^{1} + A_{22}x^{2} = b^{2}$
 $x^{1} \ge 0$
 x^{2} free,

where $A_{11} \in \mathbb{R}^{l \times k}$, $A_{12} \in \mathbb{R}^{l \times (n-k)}$, $A_{21} \in \mathbb{R}^{(m-l) \times k}$ and $A_{22} \in \mathbb{R}^{(m-l) \times (n-k)}$. The corresponding dual problem is expressed as follows, with variables $y^1 \in \mathbb{R}^l$ and $y^2 \in \mathbb{R}^{m-l}$.

$$\begin{array}{ll} (DP) & \max & w = y^1 b^1 + y^2 b^2 \\ \text{subject to} & y^1 A_{11} + y^2 A_{21} \leq c^1 \\ & y^1 A_{12} + y^2 A_{22} = c^2 \\ & y^1 \geq 0 \\ & y^2 & \text{free.} \end{array}$$
 (A.53)

It can be easily demonstrated that the two problems are mutually primal-dual and the duality theorem holds between them. Table A.2 depicts the general form of the duality relation of Linear Programming.

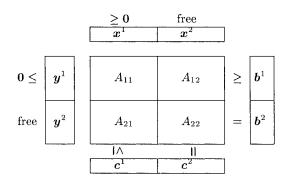


Table A.2. General Form of Duality Relation

Now, we introduce slack variables $\lambda^1 \in \mathbb{R}^l$ and $\mu^1 \in \mathbb{R}^k$ to (LP) and (DP) and rewrite them as (LP') and (DP') below:

$$(LP') \quad \min \quad z = c^{1}x^{1} + c^{2}x^{2}$$

subject to $A_{11}x^{1} + A_{12}x^{2} - \lambda^{1} = b^{1}$ (A.54)
 $A_{21}x^{1} + A_{22}x^{2} = b^{2}$
 $x^{1} \ge 0$
 x^{2} free
 $\lambda^{1} \ge 0$.

$$\begin{array}{ll} (DP') & \max & w = y^1 b^1 + y^2 b^2 \\ \text{subject to} & y^1 A_{11} + y^2 A_{21} + \mu^1 = c^1 \\ & y^1 A_{12} + y^2 A_{22} = c^2 \\ & y^1 \geq 0 \\ & y^2 & \text{free} \\ & \mu^1 \geq 0. \end{array}$$
 (A.55)

We then have the following complementarity theorem:

Corollary A.2 (Complementarity Theorem in General Form) Let (x^1, x^2, λ^1) and (y^1, y^2, μ^1) be feasible to (LP') and (DP'), respectively. Then they are optimal to (LP') and (DP') if and only if the relation below holds.

$$\boldsymbol{\mu}^1 \boldsymbol{x}^1 = \boldsymbol{y}^1 \boldsymbol{\lambda}^1 = 0. \tag{A.56}$$

Also, there exist optimal solutions that satisfy the following strong complementarity. Corollary A.3 (Strong Theorem of Complementarity) In the optimal solutions to the primal-dual pair LPs, (LP') and (DP'), there exist ones such that, in the complementarity condition (A.56), if one of the pair is 0, then the other is positive.

Notes

- 1. S.I. Gass (1985), Linear Programming, 5th ed., McGraw-Hill.
- 2. A. Charnes and W.W. Cooper (1961), Management Models and Industrial Applications of Linear Programming, (Volume 1 & 2), John Wiley & Sons.
- 3. O.L. Mangasarian (1969), Nonlinear Programming, McGraw-Hill.
- 4. K. Tone (1978), Mathematical Programming, (in Japanese) Asakura, Tokyo.
- 5. G.B. Dantzig (1963), *Linear Programming and Extensions* (Princeton: Princeton University Press).
- 6. W.A. Spivey and R.M. Thrall (1970), *Linear Optimization* (New York: Holt, Rinehart and Winston).
- 7. E.D. Nering and A.W. Tucker (1993), *Linear Programming and Related Problems* (New York: Academic Press).

Appendix B Introduction to DEA-Solver

This is an introduction and manual for the attached DEA-Solver. There are two versions of DEA-Solver, the "Learning Version" (called **DEA-Solver-LV**, in the attached CD) and the "Professional Version" (called **DEA-Solver-PRO**: visit the DEA-Solver website at: http://www.saitech-inc.com/ for further information). This manual serves both versions. DEA-Solver was developed by Kaoru Tone. All responsibility is attributed to Tone, but not to Cooper and Seiford in any dimension.

B.1 PLATFORM

The platform for this software is Microsoft Excel 97/2000 or later (a trademark of Microsoft Corporation).

B.2 INSTALLATION OF DEA-SOLVER

The accompanying installer will install DEA-Solver and sample problems in the attached CD-ROM to the hard disk (C:) of your PC. Click Setup.EXE in the folder "DEA-Solver" in the CD-ROM. Just follow the instruction on the screen. The folder in the hard disk is "C:\DEA-Solver" which includes the code DEA-Solver-LV(V3).xls and another folder "Samples(LV3)." A shortcut to DEA-Solver.xls will be automatically put on the Desktop. If you want to install "DEA-Solver" to another drive or to other folder (not to "C:\DEA-Solver"), just copy it to the disk or to the folder you designate. For the "Professional Version" an installer will automatically install "DEA-Solver."

B.3 NOTATION OF DEA MODELS

DEA-Solver applies the following notation for describing DEA models.

<Model Name> - <I or O> - <C or V or GRS>

where I or O corresponds to "Input"- or "Output"-orientation and C, V or GRS to "Constant", "Variable" or "General" returns to scale. For example, "AR-I-C" means the Input oriented Assurance Region model under Constant returns-to-scale assumption. In some cases, "I or O" and/or "C or V" are omitted. For example, "CCR-I" indicates the Input oriented CCR model which is naturally under constant returns-to-scale. "FDH" (= Free Disposal Hull) has no extensions. The abbreviated model names correspond to the following models,

- 1. CCR = Charnes-Cooper-Rhodes model (Chapters 2, 3)
- 2. BCC = Banker-Charnes-Cooper model (Chapters 4, 5)
- 3. IRS = Increasing Returns-to-Scale model (Chapter 5)
- 4. DRS = Decreasing Returns-to-Scale model (Chapter 5)

- 5. GRS = Generalized Returns-to-Scale model (Chapter 5)
- 6. AR = Assurance Region model (Chapter 6)
- 7. ARG = Assurance Region Global model (Chapter 6)
- 8. NCN = Non-controllable variable model (Chapter 7)
- 9. NDSC = Non-discretionary variable model (Chapter 7)
- 10. BND = Bounded variable model (Chapter 7)
- 11. CAT = Categorical variable model (Chapter 7)
- 12. SYS = Different Systems model (Chapter 7)
- 13. SBM = Slacks-Based Measure model (Chapter 4)
- 14. Weighted SBM = Weighted Slacks-Based Measure model (Chapter 4)
- 15. Cost = Cost efficiency model (Chapter 8)
- 16. New-Cost = New-Cost efficiency model (Chapter 8)
- 17. Revenue = Revenue efficiency model (Chapter 8)
- 18. New-Revenue = New-Revenue efficiency model (Chapter 8)
- 19. Profit = Profit efficiency model (Chapter 8)
- 20. New-Profit = New-Profit efficiency model (Chapter 8)
- 21. Ratio = Ratio efficiency model (Chapter 8)
- 22. Bilateral = Bilateral comparison model (Chapter 7)
- 23. FDH = Free Disposal Hull model (Chapter 4)
- 24. Window = Window Analysis (Chapter 9)
- 25. Super-efficiency = Super-efficiency model (Chapter 10)

B.4 INCLUDED DEA MODELS

The "Learning Version" includes all models and can solve problems with up to 50 DMUs; The "Professional Version" includes *Malmquist, Scale elasticity, Congestion* and *Undesirable output* models in addition to the above models and can deal with large-scale problems within the capacity of Excel worksheet.

B.5 PREPARATION OF THE DATA FILE

The data file should be prepared in an Excel Workbook prior to execution of DEA-Solver. The formats are as follows:

328 INTRODUCTION TO DATA ENVELOPMENT ANALYSIS AND ITS USES

B.5.1 The CCR, BCC, IRS, DRS, GRS, SBM, Super-Efficiency and FDH Models

Figure B.1 shows an example of data file for these models.

1. The first row (Row 1)

The first row (Row 1) contains Names of Problem and Input/Output Items, i.e.,

Cell A1 = Problem Name

Cell B1, C1, \ldots = Names of I/O items.

The heading (I) or (O), showing them as being input or output should head the names of I/O items. The items without an (I) or (O) heading will not be considered as inputs and outputs. The ordering of (I) and (O) items is arbitrary.

2. The second row and after

The second row contains the name of the first DMU and I/O values for the corresponding I/O items. This continues up to the last DMU.

3. The scope of data domain

A data set should be bordered by at least one blank column at right and at least one blank row at bottom. This is a necessity for knowing the scope of the data domain. The data set should start from the top-left cell (A1).

4. Data sheet name

A preferable sheet name is "DAT" (not "Sheet 1"). Never use names "Score", "Rank", "Projection", "Weight", "WeightedData", "Slack", "RTS", "Window", "Graph1" and "Graph2" for data sheet. These are reserved for this software.

The sample problem "Hospital(CCR)" in Figure B.1 has 12 DMUs with two inputs "(I)Doctor" and "(I)Nurse" and two outputs "(O)Outpatient" and "(O)Inpatient". The data set is bordered by one blank column (F) and by one blank row (14). The GRS model has the constraint $L \leq \sum_{j=1}^{n} \lambda_j \leq U$. The values of $L(\leq 1)$ and $U(\geq 1)$ must be supplied through the Message-Box on the display by request. Defaults are L = 0.8 and U = 1.2.

As noted in 1. above, items without an (I) or (O) heading will not be considered as inputs or outputs. So, if you delete "(I)" from "(I)Nurse" to "Nurse," then "Nurse" will not be accounted for in this efficiency evaluation. Thus you

	A	В	C	D	Е	F
1	Hospital	(I)Doctor	(I)Nurse	(0)Outpatient	(O)Inpatient	
2	A	20	151	100	90	
3	В	19	131	150	50	
4	С	25	160	160	55	
5	D	27	168	180	72	
6	E	22	158	94	66	
7	F	55	255	230	90	
8	G	33	235	220	88	
.9	H	31	206	152	80	
10	l	30	244	190	100	
11	J	50	268	250	100	
12	K	53	306	260	147	
13	L	38	284	250	120	
14						

can add (delete) items freely to (from) inputs and outputs without changing your data set.

Figure B.1. Sample.xls in Excel Sheet

B.5.2 The AR Model

Figure B.2 exhibits an example of data for the AR (Assurance Region) model. This problem has the same inputs and outputs as in Figure B.1. The constraints for the assurance region are described in rows 15 and 16 after "one blank row" at 14. This blank row is necessary for separating the data set and the assurance region constraints. These rows read as follows: the ratio of weights "(I)Doctor" vs. "(I)Nurse" is not less than 1 and not greater than 5 and that for "(O)Outpatient" vs. "(O)Inpatient" is not greater than 0.2 and not less than 0.5. Let the weights for Doctor and Nurse be v(1) and v(2), respectively. Then the first constraint implies

$$1 \le v(1)/v(2) \le 5.$$

Similarly, the second constraint means that the weights u(1) (for Outpatient) and u(2) (for Inpatient) satisfies the relationship

$$0.2 \le u(1)/u(2) \le 0.5.$$

	Α	В	C	D	E	F
1	Hospital	(I)Doctor	(I)Nurse	(0)Outpatient	(O)Inpatient	
2	Α	20	151	100	90	
3	В	19	131	150	50	
4	С	25	160	160	55	
5	D	27	168	180	72	
6	E	22	158	94	66	
7	F	55	255	230	90	
8	G	33	235	220	88	
9	Н	31	206	152	80	
10	Ι	30	244	190	100	
11	J	50	268	250	100	
12	К	53	306	260	147	
13	L	38	284	250	120	
14						
15	1	(I)Doctor	(I)Nurse	5		
16	0.2	(O)Outpatient	(O)Inpatient	0.5		
17						

Notice that the weights constraint can be applied between inputs and outputs as well.

Figure B.2. Sample-AR.xls in Excel Sheet

B.5.3 The ARG Model

Instead of restricting ratios of virtual multipliers, this model imposes bounds on the virtual input (output) relative to the total virtual input (output). For example, in the above hospital case, the virtual input of Doctor is expressed by $v(1) \times (Number of)$ Doctor and the total virtual input is denoted by $v(1) \times (Number of)$ Doctor + $v(2) \times (Number of)$ Nurse, where v(1) and v(2)are weights to Doctor and Nurse, respectively. We impose lower and upper bounds, L and U, to the ratio of these two factors. Thus, we have constraints as expressed below.

$$L \le \frac{v(1) \times \text{Doctor}}{v(1) \times \text{Doctor} + v(2) \times \text{Nurse}} \le U.$$

In the Excel worksheet, we designate L and U along with the input (output) name as exhibited in Figure B.3. This means that L = 0.5 and U = 0.8 for Doctor in the above expression. See Section 6.3 in Chapter 6.

	A	B	С	D	EF
	Sample-ARC	(I)Doctor	(I)Nurse	(O)Outpatien	(O)Inpatient
2	A	20	151	100	90
3	В	19	131	150	50
·	С	25	160	160	55
5	D	27	168	180	72
6	E	22	158	94	66
7	F	55	255	230	90
- 8	G	33	235	220	88
9	H	31	206	152	80
10		30	244	190	100
	J	50	268	250	100
12	K	53	306	260	147
13	L	38	284	250	120
14					
15	0.5	(I)Doctor	0.8		
16	0.2	(I)Nurse	0.3		
17	0.2	(O)Outpatier	0.5		
-18	0.4		0.8		
19					

Figure B.3. Sample-ARG.xls in Excel Sheet

B.5.4 The NCN and NDSC Models

The non-controllable and non-discretionary models have basically the same data format as the CCR model. However, the uncontrollable inputs or outputs must have the headings (IN) or (ON), respectively. Figure B.4 exhibits the case where 'Doctor' is an uncontrollable (i.e., "non-discretionary" or "exogenously fixed") input and 'Inpatient' is an uncontrollable output.

	A	В	C	- D	÷Ε	F
1	Hospital	(IN)Doctor	(I)Nurse	(0)Outpatient	(ON)Inpatient	100-041110 VE 0 100
2	Α	20	151	100	90	
3	В	19	131	150	50	
4	С	25	160	160	55	
5	D	27	168	180	72	
6	E	22	158	94	66	
7	F	55	255	230	90	
8	G	33	235	220	88	
9	н	31	206	152	80	
10	I	30	244	190	100	
- 11	J	50	268	250	100	
12	К	53	306	260	147	
13	L	38	284	250	120	
14						

Figure B.4. Sample-NCN (NDSC).xls in Excel Sheet

B.5.5 The BND Model

The bounded inputs or outputs must have the headings (IB) or (OB). Their lower and upper bounds should be designated by the columns headed by (LB) and (UB), respectively. These (LB) and (UB) columns must be inserted immediately after the corresponding (IB) or (OB) column. Figure B.5 implies that 'Doctor' and 'Inpatient' are bounded variables and their lower and upper bounds are given by the columns (LB)Doc., (UB)Doc., (LB)Inpat., and (UB)Inpat, respectively.

	A	в	C	D	E	F	G	H	1
1	Hospital	(IB)Doc.	(LB)Doc.	(UB)Doc.	(I)Nurse	(O)Outpat	(OB)Inpat.	(LB)Inpat.	(UB)Inpat.
2	A	20	15	22	151	100	90	80	100
3	В	19	15	23	131	150	50	45	55
4	C	25	20	25	160	160	55	50	60
5	D	27	21	27	168	180	72	70	76
6	E	22	20	25	158	94	66	60	80
7	F	55	45	56	255	230	90	80	100
8	G	33	31	36	235	220	88	80	95
9	Н	31	29	33	206	152	80	70	90
10	I	30	28	31	244	190	100	90	110
(1) 1	J	50	45	50	268	250	100	90	120
12	К	53	45	54	306	260	147	130	160
13	L	38	30	40	284	250	120	110	130
14									

Figure B.5. Sample-BND.xls in Excel Sheet

B.5.6 The CAT, SYS and Bilateral Models

These models have basically the same data format as the CCR model. However, in the last column they must have an integer showing their category, system or bilateral group, as follows.

For the CAT model, the number starts from 1 (DMUs under the most difficult environment or with the most severe competition), 2 (in the second group of difficulty) and so on. It is recommended that the numbers be continuously assigned starting from 1.

For the SYS model, DMUs in the same system should have the same integer starting from 1.

For the Bilateral model, DMUs must be divided into two groups, denoted by 1 or 2.

	A	В	C	D	E	F
1	Hospital	(I)Doctor	(I)Nurse	(O)Outpatient	(O)Inpatient	Cat.
2	A	20	151	100	90	1
3	В	19	131	150	50	2
4	С	25	160	160	55	2
5	D	27	168	180	72	2
6	Е	22	158	94	66	1
	F	55	255	230	90	1
8	G	33	235	220	88	2
9	Н	31	206	152	80	1
10	I	30	244	190	100	1
11	J	50	268	250	100	2
12	К	<u>5</u> 3	306	260	147	2
13	L	38	284	250	120	2
14						

Figure B.6 exhibits a sample data format for the CAT model.

Figure B.6. Sample-CAT.xls in Excel Sheet

B.5.7 The Cost and New-Cost Models

The unit cost columns must have the heading (C) followed by the *input* name. The ordering of columns is arbitrary. If an input has no cost column, its cost is regarded as zero. Figure B.7 is a sample.

	Α	В	C	D	E	F	G	Н
1	Hospital	(I)Doctor	(C)Doctor	(I)Nurse	(C)Nurse	(O)Outpat.	(O)Inpat.	
2	Α	20	500	151	100	100	90	
3	В	19	350	131	80	150	50	
4	С	25	450	160	90	160	55	
5	D	27	600	168	120	180	72	
6	E	22	300	158	70	94	66	
7	F	55	450	255	80	230	90	
8	G	33	500	235	100	220	88	
9	Н	31	450	206	85	152	80	
10	I	30	380	244	76	190	100	
11-	J	50	410	268	75	250	100	
12	К	53	440	306	80	260	147	
13	L	38	400	284	70	250	120	
14								

Figure B.7. Sample-Cost(New-Cost).xls in Excel Sheet

B.5.8 The Revenue and New-Revenue Models

The unit price columns must have the heading (P) followed by the *output* name. The ordering of columns is arbitrary. If an output has no price column, its price is regarded as zero. See Figure B.8 for an example.

	A	В	С	D	Е	F	G	Н
1	Hospital	(I)Doctor	(I)Nurse	(O)Outpat.	(P)Outpat.	(O)Inpat.	(P)Inpat.	
2	Α	20	151	100	550	90	2010	
3	В	19	131	150	400	50	1800	
4	С	25	160	160	480	55	2200	
5	D	27	168	180	600	72	3500	
6	E	22	158	94	400	66	3050	
7	F	55	255	230	430	90	3900	
8	G	33	235	220	540	88	3300	
9	Н	31	206	152	420	80	3500	
10	Ι	30	244	190	350	100	2900	
11	J	50	268	250	410	100	2600	
12	К	53	306	260	540	147	2450	
13	L	38	284	250	295	120	3000	
14								

Figure B.8. Sample-Revenue(New-Revenue).xls in Excel Sheet

B.5.9 The Profit, New-Profit and Ratio Models

As a combination of *Cost* and *Revenue* models, these models have cost columns headed by (C) for inputs and price columns headed by (P) for outputs.

B.5.10 The Window Models

Figure B.9 exhibits an example of data format for a Window Analysis model. Top-left corner (A1) contains the problem name, e.g., "Car" in this example. The next right cell (B1) must include the first time period, e.g., "89." The second row beginning from the B column exhibits "I/O items", e.g., "(I)Sales" and "(O)Profit." The name of DMUs appears from the third row in the column A. The contents (observed data) follow in the third row and after. This style is repeated until the last time period. Notice that each time period is placed at the top-left corner of the corresponding frame and (I)/(O) items have the same names throughout the time period. It is not necessary to insert headings (I)/(O) to the I/O names of the second time period and after. I/O items are determined as designated in the first time period. Figure B.9 demonstrates performance of four car-manufacturers, i.e., Toyota, Nissan, Honda and Mitsubishi, during five time periods, i.e., from (19)89 to (19)93, in terms of the input "Sales" and

the output "Profit."

A	В	С	D	E	F	G
Car	89		90		91	
DMU	(I)Sales	(O)Profit	Sales	Profit	Sales	Profit
Toyota	719	400	800	539	850	339
Nissan	358	92	401	139	418	120
Honda	264	74	275	100	280	65
Mitsubishi	190	44	203	49	231	66

	H	I	J	К
	92		93	
N	Sales	Profit	Sales	Profit
\square	894	125	903	103
$\neg \gamma$	427	34	390	0
	291	54	269	33
	255	56	262	57

Figure B.9. Sample-Window.xls in Excel Sheet

B.5.11 Weighted SBM Model

This model requires weights to inputs/outputs as data. They should be given at the rows below the main body of data set with one inserted blank row. See Figure B.10. The first column (A) has **WeightI** or **WeightO** designating input or output, respectively, and the weights to inputs or outputs follow consecutively in the order of input (output) items recorded at the top row. The values are relative, since the software normalizes them properly. Refer to (4.81)-(4.83) in Chapter 4. If they are vacant, weights are regarded as even. Figure B.10 designates that weights to Doctor and Nurse are 10:1 and those to Outpatient

	A	В	С	D	E F
1	WSBM	(I)Doctor	(I)Nurse	(O)Outpatier	(O)Inpatient
2	A	20	151	100	90
3	В	19	131	150	50
4	C	25	160	160	55
5	D	27	168	180	72
6	E	22	158	94	66
7	F	55	255	230	90
8	G	33	235	220	88
9					
10	Weightl	10	1		
11	WeightC	1	5		
12					

and Inpatient are 1:5.

Figure B.10. Sample-Weighted SBM.xls in Excel Sheet

B.6 STARTING DEA-SOLVER

After completion of the data file in an Excel sheet on an Excel book as mentioned above, close the data file and click either the icon or the file "DEA-Solver" in Explorer. This starts DEA-Solver. First, click "<u>Enable Macros</u>" and then follow the instructions on the display.

Otherwise if the file "DEA-Solver" is already open (loaded), click "Tools" on the *Menu Bar*, then select "Macro" and click "Macros." Finally, click "Run" on the Macro.

This Solver proceeds as follows,

- 1. Selection of a DEA model
- 2. Selection of a data set in Excel Worksheet
- 3. Selection of a Workbook for saving the results of computation and
- 4. DEA computation

B.7 RESULTS

The results of computation are stored in the selected Excel workbook. The following worksheets contain the results, although some models lack some of them.

1. Worksheet "Summary"

This worksheet shows statistics on data and a summary report of results obtained.

2. Worksheet "Score"

This worksheet contains the DEA-score, reference set, λ -value for each DMU in the reference set and ranking in input and in the descending order of efficiency scores. A part of a sample Worksheet "Score" is displayed in Figure B.11, where it is shown that DMUs A, B and D are efficient (Score=1) and DMU C is inefficient (Score=0.882708) with the reference set composed of B ($\lambda_B = 0.9$) and D ($\lambda_D = 0.13889$) and so on.

No.	DMU	Score	Rank	Referen	ce set (lambd	a)			
1	A	1	1	A	1				
2	В	1	1	В	1				
3	C	0.8827083	8	В	0.9	D	0.13888889		
4	D	1	1	D	1				
5	Е	0.7634995	12	A	0.5794409	В	5.72E-02	D	0.1526401
6	F	0.8347712	10	В	0.2	D	1.11111111		
7	G	0.9019608	7	A	0.2588235	В	1.29411765		
8	Н	0.7963338	11	A	0.3866921	В	1.35E-02	D	0.6183983
9	I	0.9603922	4	A	0.6470588	В	0.83529412		
10	J	0.8706468	9	D	1.3888889				
11	к	0.955098	6	A	0.86	D	0.96666667		
12	L	0.9582043	5	A	0.6470588	В	1.23529412		

Figure B.11. A Sample Score Sheet

3. Worksheet "Rank"

This worksheet contains the ranking of DMUs in the descending order of efficiency scores.

4. Worksheet "Projection"

This worksheet contains projections of each DMU onto the efficient frontier by the chosen model.

5. Worksheet "Weight"

Optimal weights v(i) and u(i) for inputs and outputs are exhibited in this worksheet. v(0) corresponds to the constraints $\sum_{j}^{n} \lambda_{j} \geq l$ and u(0) to $\sum_{j}^{n} \lambda_{j} \leq u$. In the BCC model where l = u = 1 holds, u(0) stands for the value of the dual variable for this constraint.

6. Worksheet "WeightedData"

This worksheet shows the optimal weighted I/O values, $x_{ij}v(i)$ and $y_{rj}u(r)$ for each DMU_j (for j = 1, ..., n).

7. Worksheet "Slack"

This worksheet contains the input excesses s^- and output shortfalls s^+

338 INTRODUCTION TO DATA ENVELOPMENT ANALYSIS AND ITS USES

for each DMU. In the radial models, e.g., CCR and BCC, s^- and s^+ are calculated by using the formula (3.10) for the input-oriented case. Hence, notice that the (total) input-slacks are obtained as $s^- + (1 - \theta)x_o$. In the non-radial models, e.g., SBM (Slacks-based measure), s^- and s^+ are defined via (4.48), and they indicate the total slacks of the concerned DMU.

8. Worksheet "RTS"

In case of the BCC, AR-I-V and AR-O-V models, the returns-to-scale characteristics are recorded in this worksheet. For inefficient DMUs, returns-toscale characteristics are those of the (input- or output-oriented) projected DMUs on the frontier.

9. Graphsheet "Graph1"

The bar chart of the DEA scores is exhibited in this graphsheet. This graph can be redesigned using the Graph functions of Excel.

10. Graphsheet "Graph2"

The bar chart of the DEA scores in the ascending order is exhibited in this graphsheet. A sample of Graph2 is exhibited in Figure B.12.

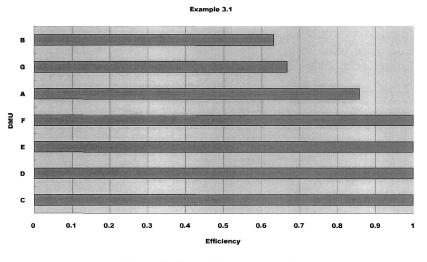


Figure B.12. A Sample Graph2

11. Worksheets "Windowk"

These sheets are only for Window models and k ranges from 1 to L (the length of time periods in the data). The contents are similar to Table 11.1 in Chapter 11. They also include two graphs, 'Variations through Window' and 'Variations by Term'. We will illustrate them in the case of Sample-Window.xsl in Figure B.9. Let k = 3 (so we deal with three adjacent years, for example). The results of computation in the case of "Window-I-C" are summarized in Table B.1.

Maker	89	90	91	92	93	Average	C Average
Toyota	0.826	1	0.592			0.806	
		1	0.592	0.208		0.600	
			1	0.351	0.286	0.546	0.651
Nissan	0.381	0.515	0.426			0.441	
		0.515	0.426	0.118		0.353	
			0.720	0.200	0	0.307	0.367
Honda	0.416	0.540	0.345			0.434	
		0.540	0.345	0.275		0.387	
			0.582	0.465	0.308	0.452	0.424
Mitsubishi	0.344	0.358	0.424	1 1 A B		0.375	
		0.358	0.424	0.326		0.369	
			0.716	0.551	0.545	0.604	0.449

 Table B.1.
 Window Analysis by Three Adjacent Years

From this table we can see row-wise averages of scores for each maker, which we call "Average through Window." The graph "Variations through Window" exhibits these averages. See Figure B.13.

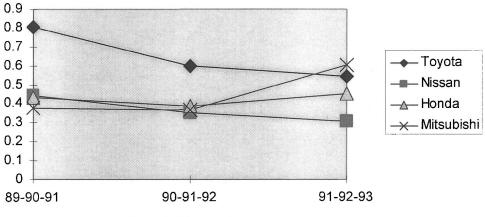
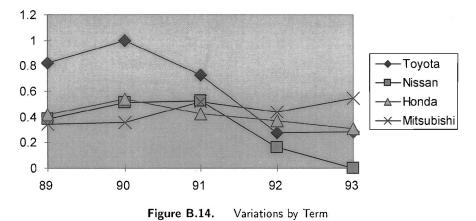


Figure B.13. Variations through Window

We can also evaluate column-wise averages of scores for each maker, which we call "Average by Term." The graph "Variations by Term" exhibits these averages. See Figure B.14.



Note. The BCC, AR-I-V and AR-O-V models contain all the worksheets except "Window." The CCR, IRS, DRS, GRS, AR-I-C, AR-O-C and SBM models contain all sheets except "RTS" and "Window." The NCN, BND, CAT, SYS, Cost, Revenue, Profit, Ratio and FDH models produce "Summary," "Rank," "Score," "Projection," "Graph1" and "Graph2." The Bilateral model shows "Summary," "Score" and "Rank" sheets. The Window models return only

B.8 DATA LIMITATIONS

"Window" and "Summary" sheets.

B.8.1 Problem Size

The "Learning Version" in the attached CD can solve problems with up to 50 DMUs. For the "Professional Version," the problem size is unlimited in terms of the number of DMUs and I/O items within the capacity of an Excel worksheet and the main memory of PC. More concretely, the data limitations for the "Professional Version" are as follows;

- 1. No. of DMUs must be less than 60000.
- 2. If No. of DMUs × (No. of Inputs + No. of Outputs + 2) \geq 60000, then the "Projection" sheet will not be provided.

B.8.2 Inappropriate Data for Each Model

DMUs with the following irregular data are excluded from the comparison group as "inappropriate" DMUs. They are listed in the Worksheet "Summary." We will adopt the following notations for this purpose.

xmax (xmin) = the max (min) input value of the DMU concerned

ymax (ymin) = the max (min) output value of the DMU concerned costmax (costmin) = the max (min) unit cost of the DMU concerned pricemax (pricemin) = the max (min) unit price of the DMU concerned

1. For the *CCR*, *BCC-I*, *IRS*, *DRS*, *GRS*, *CAT* and *SYS* models, a DMU with no positive value in inputs, i.e., $xmax \leq 0$, will be excluded from computation. Zero or minus values are permitted if there is at least one positive value in the inputs of the DMU concerned.

For the *BCC-O* model, DMUs with no positive value in outputs, i.e., $ymax \leq 0$, will be excluded from computation.

- 2. For the AR model, i.e., AR-I-C, AR-I-V, AR-O-C and AR-O-V, DMUs with $xmin < 0, xmax \le 0$ or $ymax \le 0$ will be excluded from the comparison group.
- 3. For the *FDH* model, DMUs with no positive input value, i.e., $xmax \leq 0$, or a negative input value, i.e., xmin < 0, will be excluded from computation.
- 4. For the Cost model, DMUs with $xmax \leq 0$, xmin < 0, $costmax \leq 0$, or costmin < 0 are excluded. DMUs with the current input cost ≤ 0 will also be excluded.
- 5. For the *Revenue*, *Profit* and *Ratio* models, DMUs with no positive input value, i.e., $xmax \leq 0$, no positive output value, i.e., $ymax \leq 0$, or with a negative output value, i.e., ymin < 0, will be excluded from computation. Furthermore, in the *Revenue* model, DMUs with *pricemax* ≤ 0 , or *pricemin* < 0 will be excluded from the comparison group. In the *Profit* model DMUs with $costmax \leq 0$ or costmin < 0 will be excluded. Finally, in the *Ratio* model, DMUs with $pricemax \leq 0$, $costmax \leq 0$ or costmin < 0, will be excluded.
- 6. For the NCN and BND models, negative input and output values are automatically set to zero by the program. DMUs with $xmax \leq 0$ in the controllable (discretionary) input variables will be excluded from the comparison group as "inappropriate" DMUs. In the BND model, the lower bound and the upper bound must enclose the given (observed) value, otherwise these values will be adjusted to the given value.
- 7. For the Window-I-C and Window-I-V models, no restriction exists for output data, i.e., positive, zero or negative values for outputs are permitted. However, DMUs with $xmax \leq 0$ will be characterized as being zero efficiency. This is for purpose of completing the score matrix. So, care is needed for interpreting the results in this case. If the number of DMUs per one period (term) exceeds 255, no graph will be produced.
- 8. For the *SBM* model, nonpositive inputs or outputs are replaced by a small positive value.

9. For the *Bilateral* model, we cannot compare two groups if some inputs are zero for one group while the other group has all positive values for the corresponding input item.

B.9 SAMPLE PROBLEMS AND RESULTS

The attached "DEA-Solver LV (learning version)" includes the sample problems and results for all models in the folder "Samples".

The "Professional Version" is available via http://www.saitech-inc.com.

B.10 SUMMARY OF HEADINGS TO INPUTS/OUTPUTS

Table B.2 exhibits headings to input/output and samples.

Heading	Description	Example	Models employed
(I)	Input	(I)Employee	All models
(O)	Output	(O)Sales	All models
(IN)	Non-controllable or	(IN)Population	NCN (Non-controllable)
	Non-discretionary input		NDSC (Non-discretionary)
(ON)	Non-controllable or Non-discretionary output	(ON)Area	As above
(IB)	Bounded input	(IB)Doctor	BND (Bounded variable)
(OB)	Bounded output	(OB)Attendance	As above
(LB)	Lower bound of bounded variable	(LB)Doctor	As above
(UB)	Upper bound of bounded variable	(UB)Doctor	As above
(C)	Unit cost of input	(C)Manager	Cost, New-Cost,
• •	-	., 0	Profit, New-Profit, Ratio
(P)	Unit price of output	(P)Laptop	Revenue, New-Revenue, Profit, New-Profit, Ratio

Table B.2. Headings to Inputs/Outputs

Appendix C Bibliography

Comprehensive bibliography of 2800 DEA references is available in the attached CD-ROM.

Adolphson, 228 Ahn, 91, 116, 131 Aigner, 279 Ali, xxxi, 71, 95, 107, 112, 116 Allen, 200 Andersen, 297, 301 Arnold, 63, 71, 281 Aronson, 71 Athanassopoulos, 194, 200 Banker, 60, 83, 86, 125-126, 130, 132, 153, 156, 160, 163, 203, 228 Bardhan, 63, 71, 82, 103, 125, 282 Barr, 200 Battese, 286 Baumol, 267 Bogetoft, 159 Bowlin, 108, 116 Brennan, 116 Brockett, xxvii, xxxi, 177, 181, 194, 199, 226, 228, 243 Bulla, 36 Chang, 126, 160 Charnes, 19, 21, xxviii, 33, 39, 46, 66-67, 71, 82-83, 86, 116, 131, 153, 163-165, 181, 204, 228, 272, 288, 291, 297-298, 302, 307, 315 Chu, 279 Clark, 82 Coelli, 286 Cook, xxxi, 166 Cooper, 19-21, xxviii, xxxi, 33, 35-36, 39, 46, 66-67, 71, 75, 81-83, 86, 116, 125-126, 131, 151, 153, 157-158, 160, 163-165, 181, 194, 228, 242, 261, 272, 275, 279, 286, 288, 291, 297-299, 302, 307, 315 Cummins, xxvii Dantzig, 315 Debreu, 66, 246 Deng, 211

Deprins, 107, 116 Desai, 194, 200 Dharmapala, 35, 298 Diaz, 74 Divine, 110 Dyson, 194, 199-200 Fama, xxvii, 226 Färe, 70, 102, 116, 153, 160, 246 Farkas, 320 Farrell, 33, 44, 46, 66, 69, 246, 279, 298 Ferrier, 281 Førsund, 153, 163 Fried, 71 Frisch, 163 Fukuyama, 153, 262 Gallegos, 71 Gass, 68, 315, 318 Gibson, 82 Gigerenzer, 292 Golany, xxvii, 39, 82, 166, 194, 228 Gong, 280 Gonzalez-Lima, 74 Greville, 200 Grosskopf, 70, 153, 160, 246 Haag, 297 Halek, 39 Harris, 68 Haynes, 194, 200 Hoffman, 66 Huang, 165, 181, 194, 286, 290-291, 299 Ijiri, 20 Jagannathan, 292 Jaska, 297 Jensen, xxvii, 226 Jondrow, 280 Joro, xxxi Kamakura, 228 Klopp, 39, 292 Koopmans, xxix, 45, 65, 70 Korhonen, xxxi

Korostolev, 279 Kumbhakar, 286, 299 Kwinn, 299 Land, 287 Langemeir, 194 Lee, 194 Lehman, 243 Lelas, 291 Levine, 228 Lewin, xxviii, 74, 82, 297–298 Li, 275, 279, 286, 290-291, 299 Lovell, 64, 70-71, 102, 116, 153, 246, 279, 281, 286-287, 299 Maindiratta, 114, 154, 159 Mangasarian, 315 Mann, 221 Markowitz, 289 Materov, 280 Matsui, 200 M^cCarthy, 299 Morey, 60, 74, 130, 158, 203, 228, 297 Neralic, 74, 272, 297-298 Nering, 315 Nun, 179 Olesen, 291, 299 Panzar, 267 Pareto, xxix, 45, 65 Park, xxxi, 36, 151, 261 Pastor, 102-103, 107, 112, 151, 261 Pearce, xxi Petersen, 291, 297, 299, 301 Phillips, 82 Rao, 286 Ray, 159, 301 Rhodes, 19, 21, 33, 46, 66, 71, 228 Roll, 166, 194 Rousseau, 74, 228, 242, 297 Ruefli, 110 Ruiz, 102 Russell, 102, 116 Saaty, 200 Sahoo, 158, 262 Schaible, 68 Schmidt, 71, 279 Schmitz, 39 Seiford, 20, xxviii, xxxi, 71, 74-75, 82, 95, 107, 112, 114, 116, 159–160, 163–164, 200, 275, 279, 298 Semple, 228, 242, 297 Sharpe, 289

Shen, 159 Sherman, 18, 20 Sickles, 280 Siems, 200 Simar, 107, 116, 298 Simon, xxxii, 286, 299 Singleton, 165 Sirvent, 102 Smith, 165 Spivey, 315 Storbeck, 194, 200 Stutz, 114, 164 Sueyoshi, 116, 158, 281, 293 Sugiyama, 200 Sun, 165, 194, 295 Takamura, 184 Tavares, xix Thanassoulis, 194, 199–200 Thomas, 39, 110 Thompson, 35, 74, 157–158, 163, 165, 275 Thore, 82, 286-287 Thrall, 35, 74, 107, 109, 112, 125-126, 150, 157-158, 163-165, 275, 298, 302, 315 Tone, 100, 20, 84, 96, 107, 115, 135, 148, 158, 184, 195, 200-201, 228, 243, 250, 252, 257, 259, 262, 268, 275, 315 Triantis, xxxi Tsutsui, 257, 259, 262, 268 Tsybakov, 298 Tucker, 315 Tulkens, 107, 116 Varian, 20 Wallenius, xxxi Walters, 82 Wang D, 159 Y, 242 Weber, 262 Wei, xxxi, 194 Wein, 134, 267 Whitney, 221 Wilcoxon, 221 Willig, 267 Wilson, 36, 272, 297-298, 301 Wood, 82 Yamada, 200 Yu, xxxi Zhu, 20, 74-75, 159-160, 163-164, 275, 279, 298Zi, xxvii Zionts, 71

A priori assumptions, 13 choices of weights, 13 determination, 15 knowledge, 165 Absolute bounds, 166 Academic performance, 63 Activity, 42 Activity analysis, 65 ADD-efficient, 91 Additive model, 83-84, 90, 204 non-discretionary variables, 204 vs BCC model, 92 Admissible directions, 177 directions selecting, 177 nonnegative direction vectors, 174 Advantages of DEA, 14 Agency theory, xxvii, 227 Allocative efficiency, 245-246, 260 efficiency generalization of, 166 inefficiency, xxx, 92 model, 93 AR-efficiency, 167 AR assurance region, 166, 194 efficiency, 167 Archimedean property, 71 ARG, 195 Army recruiting, xxxi, 292 Assurance region, 165-166, 178, 193, 204 Assurance region global, 195 Assurance Region Global Model, 173 Assurance region as special case of cone-ratio, 174 constraints, 171, 199 method, 166

vs cone-ratio, 174 Average behavior, 4 Average productivity, 119 Balance sheet, 179 Bank failures, 178 management incompetence, 179 merger, 147 simulation, 144 performance, 165 Banker-Morey model, 235 Banks, xxvii Japan, 144 Texas, 178 Baseball teams, 212, 241 Basel Agreement, 178 Basic Additive model, 90 Basic solution, 315-316 Basis, 315-316 Batting average, 213 BCC-efficient, 86, 88 BCC-I, 109, 161 BCC-inefficient, 88 BCC-O, 109, 161 BCC-projection, 88 BCC Banker, Charnes, Cooper model, 83 efficiency, 88 envelopment form, 89 fractional program, 87, 90 model, 83, 85 vs Additive model, 92 multiplier form, 87, 89 production possibility set, 86 reference set, 88 returns-to-scale, 124 two-phase procedure, 87 Benchmark, 4 Benchmarking, 2 Best performance, 4

Best ratio, 13 Best set of weights, 13 Bilateral comparisons, 224, 228, 241 example, 225 model, 229 BND, 229 Bootstrap method, 279 Bounded variable model, 212 Bounded non-discretionary variable, 212 projection, 214 variable, 214 model, 212, 229 Bounds absolute, 166 lower and upper, 173 on multipliers, 173 on specific variables, 178 Branch banks, xxvi Branch office example, 8 Branch store, 215 example, 3 Canada, xxvi Candidate reference point, 218 Candidate sites, 184 CAT, 229 Categorical (classificatory) variable, 204 level, 215 algorithm, 218 levels expanded, 239 model, 216 variable, xxxi, 204, 228 Categories, 240 CCP, 286, 291 CCR-AR model, 167 CCR-efficiency, 100, 24, 46 CCR-efficient, 65, 86 CCR-I, 67 CCR-inefficiency, 25 CCR-0, 67 CCR-type models, 84 CCR Charnes, Cooper, Rhodes model, 21 dual problem, 41 model, 13, 21 projection, 47, 49, 209 returns-to-scale, 126 two-phase procedure, 46 weak efficiency form, 99 Central tendency behavior, 4 Chain-store example, 19 Chance-constrained programming, xxxii, 286, 296 Charnes-Cooper-Rousseau-Semple model, 235Charnes-Cooper transformation, 69, 256

Chebychev norm, 273, 275 Cities, xxvi Classificatory variable, xxxi, 204 Cobb-Douglas, 105, 113, 280 envelopment, 114 form, 281 regression, 281 Column views, 293 Combustion engineering, 6 Commercial banks, 194 Comparing DEA scores of two groups, 223 efficient frontiers of two groups, 224 Comparisons of different systems, 219 of efficiencies, 221 of efficiency, 229 Complementarity theorem, 319 Complementary condition, 46 Complementary slackness, xxix, 81, 172 and sensitivity analysis, 75 Composed error, 279 Computational aspects of DEA, 50 Computational procedures, 50 Conditional chance constraints, 291 Cone-ratio, 165 as generalization of assurance region, 174 envelopment, 165, 178, 204 example, 176 method, 174, 193 formula, 175 vs assurance region, 174 Confidence intervals, 279 Consensus-making, 184 Consensus formation, 184 Consistent estimator, 279 Constant returns-to-scale, 83, 121, 126 assumption, 4, 42, 64, 83 Contestable markets, 267 Controllable categorical level, 217 variable, 207 Convex cone, 174 hull, 83 Convexity assumption, 219 condition, 87, 112, 138 Cost-C(V), 262Cost efficiency, 247, 262 Countries, xxvi CRS, 136 Cyprus, xxvi Data Envelopment Analysis, 2, 33, 52 Data file, 327 Data

imprecise, xxxiii inappropriate, 340 limitation, 340 variability, 296 variations, 271 DEA-Solver-LV, 326 DEA-Solver-PRO, 326 DEA-Solver, 17, xxix, 42, 53, 67, 109, 161, 194, 228, 238, 262, 311, 326 AR model, 329 Bilateral model, 332 BND model, 332 CAT model, 332 Cost model, 333 data file, 327 inappropriate data, 340 installation, 326 learning version, 326-327 limitations, 340 manual, 326 model notation, 326 NCN model, 331 NDSC model, 331 New-Cost model, 333 New-Profit model, 334 New-Revenue model, 334 professional version, xxx, 326-327 Profit model, 334 Ratio model, 334 results, 336 Revenue model, 334 sample problems, 342 starting, 336 SYS model, 332 website, 326 Window model, 334 worksheets, 336 DEA, 2 DEA likelihood estimator, 279 DEA web sites, xxviii Decision Making Unit, 22, xxvii, 33 Decision Sciences Institute, xxviii Decomposition of efficiency, 141 of technical efficiency, 140, 143 Decreasing returns-to-scale, 83, 121, 139 Definition of SBM, 96 Degrees of discretion, 204 Degrees of freedom, 271–272 Demutualizations, xxvii Department stores, 221 Deteriorating behavior, 293 Deterministic, 287 Diet of Japan, 184 Different organization forms, 226 Dimension free, 96 Discretionary variable, 60, 204, 292

Distance measure, 9 DMU, 22, 33 DRS-I, 161 DRS-0, 161 DRS, 136 DSI-Athens meeting, xxviii Dual feasible solution, 317 Dual problem, 43, 317 Dual simplex, 52 Dual theorem of linear programming, 197 Duality, 315 relations, xxix theorem, 35, 317, 319 Dummy variables, xxxi, 281 Economic efficiency, 11 Economics, xxviii Economists, xxviii Effectiveness, 63 vs efficiency, 179 Effects of a merger, 147 Efficiency, 3, 63 audits, 110 dominance, 291 prices, 66 variance, 12 vs effectiveness, 179 Efficient frontier, 3, 25, 85, 90, 127 Elasticity, 121, 133 of scale, 121 Electric power industries, 221 England, xxvi Envelop, 3 Enveloped, 52 Envelopment form, xxix, 52, 89 map, 37 model, 65 Equal sample size, 243 Euclidean measure, 10 Evaluation by experts, 179 Excel data sheet, xxix Exogenously fixed, 60 inputs and outputs, 228 Expert evaluations, 179 Extended additive model, 206 Extreme efficient point, 277 Facet, 127 Farrell efficiency, 44, 46, 69 Fast-food outlets, xxxi Fast-food restaurants, 60 FDH, 107, 110, 242 mixed integer program, 108 model, 85 Feasible solution, 315 Fighter Command of U.S. Air Force, 60 Financial institutions, 144

Fixed weights, 12, 14 vs. variable weights, 12 Formula for cone-ratio method, 175 Fractional programming, 69, 256 formulation for BCC, 90 problem, 23, 268 program, 24 France, xxvi Free disposal, 70 Free disposal hull example, 107 model, 85, 107 Functional efficiency, 66 Furnace efficiency, 6 Fuzzy sets, xxxi Gauge of scale efficiency, 130 General hospital case, 176 example, 169 General returns-to-scale, 109 Generalization of allocative and price efficiency, 166 Generalized inverse, 200 Generalized returns-to-scale, 139 Global technical efficiency, 140 Groups of DMUs, 223 GRS-I, 161 GRS-O, 161 GRS, 109, 139 Halfspaces, 123 Hierarchical category, 215 model, 229 High-energy physics laboratory, 165 Home runs, 213 Hospital example, 12, 162 Hyperplane, 34, 123 Imprecise Data, xxxiii Imputed (shadow) price, 172 Inappropriate data, 340 Income statement, 179 Inconsistent constraints, 286 Incorporating judgement, 165 Increasing returns-to-scale, 83, 121, 138 Infeasibility, 310 Input-oriented, 105 Input-oriented non-radial super-efficiency model, 308 Input-oriented assurance region model, 194 BCC model, 87 CCR model, 41, 58 SBM model, 142 Input data matrix, 22 excess, 44 mix inefficiency, 97

numeraire, 166 reduction, 92 weights, 23 Insurance, 182 companies, xxvi industry, 226 Inter-group comparison, 224 Interior point algorithm, 279 IRS-I, 161 IRS-0, 161 IRS, 136 Isoquant, 246 Japanese banks, 144 Japanese capital, 184 Japanese Financial Supervisory Agency, 147 Jet Engines, 36 Joint chance constraints, 291 Joint constraints, 291 K-Efficiency, 66 Kinds of efficiency, xxx Learning version of DEA-Solver, 326 Levels of service, 217 Library problem, 225 Linear program, 23 Linear programming, xxix, 315 problem, 315 Linkage constraints, 197 Linked-cone DEA, 197 Linked cone, 197 Linked constraint, 193 Local pure technical efficiency, 140 Log-linear, 280 Lower and upper bounds, 172 Maintenance activities, xxvi Management Science, xxviii Management scientists, xxix Managerial efficiency, 66 performance, 179 Mann-Whitney statistic, 243 Marginal productivity, 119 Massachusetts, 18, xxviii Max-slack solution, 45, 47, 49 Maximizing profit, 256 Maximizing revenue/expenses ratio, 256 Maximum capacity, 212 MCDM, xxxi MDI (Minimum Discrimination Information) Statistic, xxvii Merger effect, 147 Methodological bias, 281 Minimal cost, 247 Mix changes, 267 efficiency, 142 inefficiencies, xxx, 45

inefficiency, 10 Model selection, 104 Monotone, 96 Most favorable weights, 25 Most productive scale size, 130, 132, 140 MPSS, 130 Moving averages, xxxii MPSS, 130, 132, 134, 150 fractional program, 132 Multicriteria decision-making, xxxi Multiple interacting objectives, 227 Multiple objective programming, xxxi Multiple optimal solutions, 98 Multiple reference sets, 81 Multiplicative efficiency, 114 Multiplicative model, 85, 113, 242 Multiplier bounds, 166 form, xxix, 52, 87, 89 model, 35, 65, 276 model approach, 275 space, 30, 34 Mutual companies, 226 Mutual funds, xxvii NCN-projection, 209 NCN, 208, 228 NDRS, 138 NDSC, 229 Negative slack values, 182 New-Cost, 263 New-Profit, 264 New-Revenue, 263 New capital, 184 New capital for Japan, 184 New insights, xxvi NIRS, 139 Non-Archimedean condition, 72 element, 71, 126, 276 infinitesimal, 71 Non-controllable variable, 207–208 model, 228 Non-decreasing returns-to-scale, 138 Non-discretionary, xxxi, 228 Non-discretionary variable model, 210 Non-discretionary variable model, 229 Non-discretionary extensions, 212 inputs and outputs, 212 variable, 60, 203, 236, 292 Non-increasing returns-to-scale, 139 Non-radial super-efficiency models, 305 Nonconvex (staircase) production possibility set, 85

Nonnegativity constraint, 23, 184 Nonnegativity requirements, 183 Nonparametric statistical tests, 222 Nonparametric statistics, 221 Nonunique reference set, 98 Nonzero slack, 204 Normalization condition, 123 Null hypothesis, 223 Number of DMUs, 106, 295 Numeraire, 166 Odense University, xxviii OLS, 281 Operations Research, xxviii Operations researchers, xxviii Optimal simplex tableau, 51 solution, 315 weights, 25 Ordinal relations, xxxi Ordinary least squares, 281 Organization forms, xxvi Output-input ratio, 1 Output-oriented, 105 Output-oriented non-radial super-efficiency model, 308 Output-oriented assurance region model, 169, 195 BCC model, 89 CCR model, 41, 59 model, 58 scale efficiency, 142 Output-to-input ratio, 1 Output data matrix, 22 expansion, 92 mix inefficiency, 97 multiplier bounds, 173 numeraire, 166 shortfall, 44 weights, 23 Overall efficiency, 247, 260 Pareto-Koopmans efficiency, xxix, 45-46 PARN, xxviii Partial productivity measure, 1 Peer (comparison) group, xxvii Peer group, 25 Phase II extension, 44 Phase II objective, 51 Physical variance, 11 Piecewise linear function, 122 Point-to-hyperplane correspondence, 34 Points at infinity, 46 Police forces, xxvi Polyhedral convex cone, 174 Portfolio analysis, 289 Positive data set assumption, 41

Positive threshold input value, 122 Positivity assumptions, 14 Price efficiency, 247 efficiency generalization of, 166 inefficiencies, 92 Pricing vector, 52 Primal-dual correspondences, 43, 87 Primal feasible solution, 317 problem, 317 Principal components analysis, 213 Probabilistic formulation, 271, 296 Product of Input and Output Inefficiencies, viii, 97 Production economics, xxix Production function, 119, 280 Production possibility set, 7, 41-42, 64, 85-86, 90, 119, 123, 219, 246 Productivity, 1, 3 Productivity Analysis Research Network (PARN), xxviii Professional version of DEA-Solver, xxx, 326 - 327Profit-C(V), 263 Profit-maximization mix, 248 Profit efficiency, 248, 263 Profit maximization, 267 Program efficiency, 66 Program Follow Through, 66 experiment, 33 Projection Additive model, 91 BCC, 88 CCR, 47 MPSS, 130 AR, 168 bounded, 214 NCN, 209 SBM, 98 Proportional change, 133 Public interest, 227 Public libraries, 208, 215 Public programs, 261 Public schools, 281 Public vs private universities, 116 Pure technial efficiency, 141 Purely technical inefficiency, 11, 93 Quality-of-life, xxvi Radial (proportional) efficiency, 84 efficiency, 45, 84 measure, 9, 11 Radius of stability, 274 Rank-sum-test, 221 Rank order statistic, xxvii

Ratio-C(V), 264 Ratio (revenue/cost) efficiency, 264 Ratio form, 1 Ratio form of DEA, 67 Ratio of ratios, 35 Reciprocal relation, 86 Reference set, 7, 25, 47, 98 Refusal to play, 291 Regions, xxvi Regression, xxxi Regression analysis vs DEA, 4 Regression line, 4 Regression with dummy variable values, 283 Regulated Electric Cooperatives, 110 Relative efficiency, 5 Restricted problem, 51 Restricted tableau, 51 Returns-to-scale BCC characterization, 124–125 CCR characterization, 126 characterization, 124 constant, 83 decreasing, 83 efficiency, 129 generalized, 139 identifying, 136-137 increasing, 83 inefficiencies, xxx non-decreasing, 138 non-increasing, 139 reference set characterization, 136 variable, 83 Returns to scale, 119, 150, 267 Revenue-C(V), 263 Revenue efficiency, 248 Revenue maximization, 263 Revenue vs. cost ratio, 256 Risk adjusted capital, 178 Risk coverage, 178 Robustness, 271 Round robin procedure, 293 Row views, 293 Russell input measure, 101 Russell measure, 102 Sales environment, 215 Sample size, 243 Satisficing, xxxii, 286 behavior, 286 SBM-efficiency, 100 SBM-efficient, 98 SBM-inefficient DMU, 98 SBM-projection, 98 SBM, 96, 109, 142, 293 definition of, 96 Scale efficiency, 130, 140 Scale inefficiency, 92 Schools

Texas, 203 SCSC, 279 Second stage optimization, 93 Security banks, xxvii Semipositive, 42, 162 data, 41 Semipositivity, 64 Sensitivity, xxxii, 116, 271, 296 Sensitivity analysis, xxxii, 271-272, 275 Separate efficiency frontiers, xxvii SF, 281 Shape of the Production Possibility Set, 104 Sharper discrimination among DMUs, 106 Shortage of outpatients, 169 Simple ratio definition, xxix Simplex criterion, 316 method, 52, 316 multiplier, 52, 316 Simulation experiment, 281 Single-output-to-single-input ratio, 2 Site selection, 165, 194 Sites for new capital of Japan, 184 Slack vectors, 44 Slacks-based measure of efficiency, 84, 95, 107, 142 Slacks, 183 Solvency, 243 Splitting variables, 288 Stability, xxxii, 271 analysis, 296 Staircase (or step) function, 108 Standard normal distribution, 223 State-mandated Excellence Standards for Texas, 63 Statistical approach, 279 characterizations, 271 issues, 221 methods, 279 regression, 292, 296 regression model, 280 tests of significance, 279 Status quo, 8 Steady behavior, 293 Steel industry, 267 Stochastic frontier regression, xxxiii Stochastic characterizations, 271 efficiency, 290-291 frontier, 268, 281 regression, 279 Stock companies, 226 Stock vs. mutual, xxvii Strong complementarity theorem, 46

Strong Complementary Slackness Condition, 279Strong disposal, 70 Strong theorem of complementarity, 321, 46 Summary of model characteristics, 104 of the Basic DEA Models, 104 Super-Radial, 311 Super-SBM, 312 Supercolliding Super Conductor, 194 Supermarket, 215, 221 example, 6, 240 Supporting hyperplane, 123 as dual solution, 124 nonunique, 124 Surgical units, 18 Surplus in doctors, 169 Symmetric primal-dual problems, 319 SYS, 229 Target entity, 13 TDT measure, 35 Teaching hospitals, 18 Team power, 213 Technical efficiency, 11, 45, 93, 141, 246-247 inefficiency, 10, xxx, 92 vs mix inefficiency, 93 Texas Assessment of Academic Skills (TAAS) Test, 63 Texas Banking Commission, 177 Texas banks, 178 Texas Education Agency, 62 Texas Public Utility Commission, 110 Texas schools, 203 Texas State Auditor's Office, 182 Theorem of the alternative, 320 Time series, 292 Total factor productivity, 1, 14 Transformed data, 181 Translation invariance, 93, 106-107 Trends, 296 Two-phase LP problem, 44 Two-phase procedure, 71, 87 Two-stage analysis, xxvii Two-stage DEA-regression approach, 281 Two-stage process DEA and regression, xxxi Two-stage solution procedure, 126 Unequal sample size, 243 Unit ball, 274 Unit cost, 172 United States, xxvi Unitized axes, 6 Units invariance, 5, 15 theorem, 24 Units invariant, 6, 24, 96, 114 Universities, xxvi

public vs. private, 116 University of Massachusetts-Amherst, xxviii University of Warwick, xxviii Unstable point, 274 USAREC, 292 Variable returns-to-scale, 83 Variable weights, 13–14 vs. fixed weights, 12 Vector of multipliers, 51 Virtual input, 21, 23, 33 Virtual output, 21, 23, 33 Visual Basic, xxix Wales, xxvi Warwick, xxviii Waste, 11 Weak and strong disposal assumptions, 70 Weak disposal, 70 Weak efficiency, 45-46, 84, 93,

Weak efficiency, 70 form for CCR model, 99 Weather conditions, xxxi Web site UMass, xxviii Warwick, xxviii Web sites for DEA, xxviii Website for DEA-Solver, 326 Welfare economics, xxix, 65 Welfare efficiency, xxix Wilcoxon-Mann-Whitney, x, 222 statistic, 228 Window analysis, xxxii, 271-272, 296 column views, 293 row views, 293 Wise men, 184 Within group comparison, 224 Zero order decision rules, 291