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**Report n. 243**

**Pseudoconvexity under the Charnes-Cooper  
transformation**

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**Pisa, Novembre 2003**

**- Stampato in Proprio -**

# Pseudoconvexity under the Charnes-Cooper transformation

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## Abstract

In [7] Charnes and Cooper reduce a linear fractional program to a linear program with help of a suitable transformation of variables. We show that this transformation preserves pseudoconvexity of a differentiable function.

**KeyWords** Fractional programming, pseudoconvexity, Charnes-Cooper type transformation.

2000 Mathematics Subject Classification 90C32, 26B25

## 1 Introduction

In this note we will show that the Charnes-Cooper transformation [7] as well as a generalized version of it preserve the pseudoconvexity of a differentiable function  $f$ .

Such a result generalizes the previous one given by the authors in the case where  $f$  is a twice differentiable function [16]. The result will be derived by showing that a pseudomonotone gradient of a differentiable function turns into a pseudomonotone gradient of the transformed function where the Charnes-Cooper type transformation is applied to the function ( not to the gradient). This implies the result of the paper since pseudomonotonicity of the gradient corresponds to pseudoconvexity of a differentiable function [17].

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## 2 The Charnes-Cooper transformation

Consider the following transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^n$

$$y = \frac{Ax}{b^T x + b_0} \quad (2.1)$$

defined on the set  $\Gamma = \{x \in \mathbb{R}^n : b^T x + b_0 > 0\}$  where  $A$  is a nonsingular matrix of order  $n$ ,  $b \in \mathbb{R}^n$  and  $b_0 \neq 0$ .

**Lemma 2.1** *The inverse of the transformation (2.1) is given by*

$$x = \frac{b_0 A^{-1} y}{1 - b^T A^{-1} y} \quad (2.2)$$

$$y \in \Gamma^* = \{y \in \mathbb{R}^n : \frac{b_0}{1 - b^T A^{-1} y} > 0\} \quad (2.3)$$

*Proof* From (2.1) we have  $A^{-1}y = \frac{x}{b^T x + b_0}$ ,  $b^T A^{-1}y = \frac{b^T x}{b^T x + b_0} = 1 - \frac{b_0}{b^T x + b_0}$ , so that

$$\frac{1}{b^T x + b_0} = \frac{1 - b^T A^{-1} y}{b_0} \quad (2.4)$$

The thesis is achieved taking into account that  $b^T x + b_0 > 0$  implies  $\frac{b_0}{1 - b^T A^{-1} y} > 0$  and that  $x = (b^T x + b_0) A^{-1} y$ .  $\square$

Denote with  $J_x$ ,  $J_y$  the Jacobian matrices of the transformation (2.1), (2.2), respectively; from differential calculus rules, we have

$$J_x = \frac{A}{b^T x + b_0} \left[ I - \frac{x b^T}{b^T x + b_0} \right] \quad (2.5)$$

$$J_y = \frac{b_0 A^{-1}}{1 - b^T A^{-1} y} \left[ I + \frac{y b^T A^{-1}}{1 - b^T A^{-1} y} \right] \quad (2.6)$$

Let  $J_{x(y)}$  ( $J_{y(x)}$ ) the Jacobian matrix  $J_x$  ( $J_y$ ) evaluated at  $x = \frac{b_0 A^{-1} y}{1 - b^T A^{-1} y}$  ( $y = \frac{Ax}{b^T x + b_0}$ ).

The following lemma holds.

**Lemma 2.2**

i)

$$J_{x(y)} = \frac{1 - b^T A^{-1} y}{b_0} [A - y b^T] \quad (2.7)$$

$$J_{y(x)} = (b^T x + b_0) \left[ A^{-1} + \frac{x b^T A^{-1}}{b_0} \right] \quad (2.8)$$

ii)  $J_y J_{x(y)} = I$ ,  $J_x J_{y(x)} = I$ .

Proof

i) (2.7) follows directly from (2.5) substituting (2.4) and (2.2). In a similar way (2.8) follows.

ii)  $J_y J_{x(y)} = \frac{b_0 A^{-1}}{1 - b^T A^{-1} y} \left[ I + \frac{y b^T A^{-1}}{1 - b^T A^{-1} y} \right] \cdot \frac{1 - b^T A^{-1} y}{b_0} [A - y b^T] =$   
 $= [A^{-1} + \frac{A^{-1} y b^T A^{-1}}{1 - b^T A^{-1} y}] \cdot [A - y b^T] = I - A^{-1} y b^T + \frac{A^{-1} y b^T}{1 - b^T A^{-1} y} - \frac{A^{-1} y b^T A^{-1} y b^T}{1 - b^T A^{-1} y} =$   
 $= I - A^{-1} y b^T + \frac{A^{-1} y b^T}{1 - b^T A^{-1} y} - \frac{A^{-1} y b^T (b^T A^{-1} y)}{1 - b^T A^{-1} y} = I - A^{-1} y b^T + \frac{(1 - b^T A^{-1} y) A^{-1} y b^T}{1 - b^T A^{-1} y} = I.$   
 In an analogous way it can be proved that  $J_x J_{y(x)} = I$ .  $\square$

In the following we will utilize the following lemma.

**Lemma 2.3** Let  $x, \bar{x} \in \Gamma$ ,  $y, \bar{y} \in \Gamma^*$  such that  $x = \frac{b_0 A^{-1} y}{1 - b^T A^{-1} y}$ ,  $\bar{x} = \frac{b_0 A^{-1} \bar{y}}{1 - b^T A^{-1} \bar{y}}$ . We have

$$(\bar{x} - x)^T J_{x(y)}^T = \frac{1 - b^T A^{-1} y}{1 - b^T A^{-1} \bar{y}} (\bar{y} - y)^T \quad (2.9)$$

$$(\bar{x} - x)^T = \frac{1 - b^T A^{-1} y}{1 - b^T A^{-1} \bar{y}} (\bar{y} - y)^T J_y^T \quad (2.10)$$

$$(\bar{x} - x)^T J_{x(\bar{y})}^T = \frac{1 - b^T A^{-1} \bar{y}}{1 - b^T A^{-1} y} (\bar{y} - y)^T \quad (2.11)$$

$$(\bar{x} - x)^T = \frac{1 - b^T A^{-1} \bar{y}}{1 - b^T A^{-1} y} (\bar{y} - y)^T J_{\bar{y}}^T. \quad (2.12)$$

Proof We have  $(\bar{x} - x)^T J_{x(y)}^T = \bar{x}^T J_{x(y)}^T - x^T J_{x(y)}^T$ . On the other hand:

$$- x^T J_{x(y)}^T = \frac{b_0 y^T (A^{-1})^T}{1 - b^T A^{-1} y} \left( \frac{1 - b^T A^{-1} y}{b_0} \right) (A^T - b y^T) = (1 - b^T A^{-1} y) y^T.$$

$$- \bar{x}^T J_{x(y)}^T = \frac{b_0 \bar{y}^T (A^{-1})^T}{1 - b^T A^{-1} \bar{y}} \left( \frac{1 - b^T A^{-1} y}{b_0} \right) (A^T - b y^T) = \frac{1 - b^T A^{-1} y}{1 - b^T A^{-1} \bar{y}} [\bar{y}^T - \bar{y}^T (A^{-1})^T b y^T].$$

$$\text{Consequently } (\bar{x} - x)^T J_{x(y)}^T = \frac{1 - b^T A^{-1} y}{1 - b^T A^{-1} \bar{y}} [\bar{y}^T - \bar{y}^T (A^{-1})^T b y^T - (1 - b^T A^{-1} \bar{y}) y^T] =$$

$$\frac{1 - b^T A^{-1} y}{1 - b^T A^{-1} \bar{y}} [\bar{y}^T - y^T - \bar{y}^T (A^{-1})^T b y^T + b^T A^{-1} \bar{y} y^T].$$

Since  $b^T A^{-1} \bar{y} = \bar{y}^T (A^{-1})^T b$ , (2.9) is achieved.

Taking into account that the inverse of  $J_{x(y)}$  is  $J_y$ , (2.10) follows.

From (2.9), substituting  $\bar{x}$  with  $x$  and  $x$  with  $\bar{x}$ , we obtain  $(x - \bar{x})^T J_{x(\bar{y})}^T = \frac{1 - b^T A^{-1} \bar{y}}{1 - b^T A^{-1} y} (y - \bar{y})^T$ , which is equivalent to (2.11). Finally, (2.12) follows taking into account that  $J_{x(\bar{y})}$  is the inverse of  $J_{\bar{y}}$ .  $\square$

### 3 Pseudoconvexity under the Charnes-Cooper transformation

Let  $f(x)$  be a differentiable real-valued function defined on an open and convex subset  $S$  of  $\mathbb{R}^n$  and consider the function  $\psi(y)$  obtained applying the previous Charnes-Cooper transformation to  $f(x)$ . Obviously we have  $f(x(y)) = \psi(y)$  and  $f(x) = \psi(y(x))$ .

By the differential calculus rules we have

$$\nabla f(x) = J_x^T \nabla \psi(y(x)), \quad \nabla \psi(y) = J_y^T \nabla f(x(y)).$$

We recall that a function  $F$  is pseudomonotone on an open and convex set  $C \subseteq \mathbb{R}^n$  if and only if for every pair of distinct points  $v, w \in C$  we have

$$(w - v)^T F(v) > 0 \Rightarrow (w - v)^T F(w) > 0. \quad (3.1)$$

The following theorem points out that the Charnes-Cooper transformation preserves the pseudomonotonicity of the gradient of the function.

**Theorem 3.1**  $\nabla f(x)$  is pseudomonotone if and only if  $\nabla \psi(y)$  is pseudomonotone.

*Proof* Assume the pseudomonotonicity of the gradient of  $f(x)$ . We must show that  $(\bar{y} - y)^T \nabla \psi(y) > 0$  implies  $(\bar{y} - y)^T \nabla \psi(\bar{y}) > 0$ . Taking into account Lemma (2.3) and ii) of Lemma (2.2), we have

$(\bar{y} - y)^T \nabla \psi(y) = (\bar{y} - y)^T J_y^T \nabla f(x(y)) = \frac{1-b^T A^{-1} \bar{y}}{1-b^T A^{-1} y} (\bar{x} - x)^T J_{x(y)}^T J_y^T \nabla f(x) =$   
 $= \frac{1-b^T A^{-1} \bar{y}}{1-b^T A^{-1} y} (\bar{x} - x)^T \nabla f(x)$ . Since  $\frac{1-b^T A^{-1} \bar{y}}{1-b^T A^{-1} y} > 0$ , the inequality  $(\bar{y} - y)^T \nabla \psi(y) > 0$  implies  $(\bar{x} - x)^T \nabla f(x) > 0$  so that, for the pseudomonotonicity of  $\nabla f(x)$ , we have  $(\bar{x} - x)^T \nabla f(\bar{x}) > 0$ . From (2.12) we have

$(\bar{x} - x)^T \nabla f(\bar{x}) = \frac{1-b^T A^{-1} \bar{y}}{1-b^T A^{-1} y} (\bar{y} - y)^T J_{\bar{y}}^T \nabla f(y(\bar{x})) = \frac{1-b^T A^{-1} \bar{y}}{1-b^T A^{-1} y} (\bar{y} - y)^T \nabla \psi(\bar{y})$ , so that  $(\bar{x} - x)^T \nabla f(\bar{x}) > 0$  implies  $(\bar{y} - y)^T \nabla \psi(\bar{y}) > 0$  and the thesis follows.

Assume now the pseudomonotonicity of the gradient of  $\psi(y)$ . We must prove that  $(\bar{x} - x)^T \nabla f(x) > 0$  implies  $(\bar{x} - x)^T \nabla f(\bar{x}) > 0$ . Taking into account (2.10), we have

$(\bar{x} - x)^T \nabla f(x) = \frac{1-b^T A^{-1} y}{1-b^T A^{-1} \bar{y}} (\bar{y} - y)^T J_y^T \nabla f(y(x)) = \frac{1-b^T A^{-1} y}{1-b^T A^{-1} \bar{y}} (\bar{y} - y)^T \nabla \psi(y)$  and thus  $(\bar{x} - x)^T \nabla f(x) > 0$  implies  $(\bar{y} - y)^T \nabla \psi(y) > 0$ ; for the pseudomonotonicity of  $\nabla \psi(y)$ , we have  $(\bar{y} - y)^T \nabla \psi(\bar{y}) > 0$ . Taking into account (2.11) we have

$$(\bar{y} - y)^T \nabla \psi(\bar{y}) = \frac{1-b^T A^{-1} y}{1-b^T A^{-1} \bar{y}} (\bar{x} - x)^T J_{x(\bar{y})}^T \nabla \psi(\bar{y}) = \frac{1-b^T A^{-1} y}{1-b^T A^{-1} \bar{y}} (\bar{x} - x)^T J_{x(\bar{y})}^T J_{\bar{y}}^T \nabla f(\bar{x}) = \frac{1-b^T A^{-1} y}{1-b^T A^{-1} \bar{y}} (\bar{x} - x)^T \nabla f(\bar{x}).$$

Consequently,  $(\bar{y} - y)^T \nabla \psi(\bar{y}) > 0$  implies

$(\bar{x} - x)^T \nabla f(\bar{x}) > 0$  and the thesis is complete.

□

**Corollary 3.1** *The Charnes-Cooper transformation (2.1) preserves the pseudoconvexity of a function  $f$ , that is  $f(x)$  is pseudoconvex if and only if  $\psi(y)$  is pseudoconvex.*

*Proof* It is sufficient to note that a function is pseudoconvex if and only if its gradient map is pseudomonotone.

**Remark 3.1** *While the pseudoconvexity of a function is preserved under a Charnes-Cooper transformation, the pseudomonotonicity of a map is not preserved under a Charnes-Cooper transformation. Consider for instance, the pseudomonotone function  $f(x) = x^3$ . For  $y = -\frac{x}{x+1}$  ( $x > -1$ ), hence  $x = -\frac{y}{y+1}$  ( $y > -1$ ), we obtain  $f(x(y)) = (-\frac{y}{y+1})^3$  which is not pseudomonotone. To see this, let  $y = -\frac{1}{2}$ ,  $z = 1$ . Then  $(z - y)f(x(y)) = \frac{3}{2} > 0$ , but  $(z - y)f(x(z)) = -\frac{3}{16} < 0$ .*

*To avoid a misinterpretation of Theorem 3.1, we emphasize that the Charnes-Cooper transformation is applied to the function  $f(x)$ , not to its gradient. In other words, the pseudomonotonicity of the gradient map is just a means to obtain the main result given in Corollary 3.1.*

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