

## Note

### On Invex Sets and Preinvex Functions

S. R. MOHAN AND S. K. NEOGY

*Indian Statistical Institute, 7, S.J.S. Sansanwal Marg, New Delhi, 110016, India*

*Submitted by E. Stanley Lee*

Received August 10, 1993

In this paper we consider the class of the invex function introduced by Hanson. We show that under certain condition an invex function defined on an invex set  $A$  is preinvex on  $A$ . Similarly, a quasiinvex function defined on an invex set  $A$  is prequasiinvex. © 1995 Academic Press, Inc.

#### 1. INTRODUCTION

In 1981 Hanson [3] introduced a class of functions with convex like property. This class is defined as follows.

**DEFINITION 1.1.** We say that a differentiable function  $f : R^n \rightarrow R$  belongs to Hanson's class (or satisfies Hanson's condition) if there exists a function  $\eta : R^n \times R^n \rightarrow R^n$  such that for any  $x, y \in R^n$

$$f(x) - f(y) \geq (\eta(x, y))^t \nabla f(y), \quad (1.1)$$

where  $\nabla f(y)$  is the gradient vector of  $f$  at  $y$  and for any column vector  $a$ ,  $a^t$  denotes its row transpose.

The importance of Hanson's class of functions in mathematical programming is due to the following theorem observed by Hanson [3].

**THEOREM 1.1.** *Minimize  $f(x)$  subject to*

$$g_i(x) \leq 0, \quad 1 \leq i \leq m, \quad (1.2)$$

where  $f, g_i : R^n \rightarrow R, 1 \leq i \leq m$ , are once differentiable functions. Let  $\bar{x} \in S = \{x \mid g_i(x) \leq 0, 1 \leq i \leq m\}$  and let  $\bar{x}$  satisfy the Karush Kuhn Tucker conditions [5, 6] of optimality. Then  $\bar{x}$  is a minimizer of  $f$  over  $S$ .

The name invex with respect to  $\eta$  (a short form for invariant convex) has been given to a function satisfying the Hanson property with the function  $\eta$  by Craven [2]. This is because, if  $\phi : R^n \rightarrow R^n$  is a differentiable and invertible transformation  $f$  satisfies the Hanson property with  $\eta$  iff  $f \circ \phi$  satisfies the Hanson property with  $\bar{\eta}(x, y) = J_{\phi}^{-1} \eta(\phi(x), \phi(y))$ , where  $J_{\phi}^{-1}$  denotes the Jacobian of  $\phi^{-1}$ .

**DEFINITION 1.2.** Let  $f : R^n \rightarrow R$ . We say that  $f$  is pseudoinvex with respect to  $\eta : R^n \times R^n \rightarrow R^n$  if

$$(\eta(x, y))' \nabla f(y) \geq 0 \Rightarrow f(x) \geq f(y).$$

**DEFINITION 1.3.** Let  $f : R^n \rightarrow R$  be differentiable. We say that  $f$  is quasiinvex with respect to  $\eta$ , where  $\eta : R^n \times R^n \rightarrow R^n$ , if

$$f(x) \leq f(y) \Rightarrow (\eta(x, y))' \nabla f(y) \leq 0.$$

In this note we study invex sets. Although there are examples of such sets in the literature, they are mostly in  $R$ . See Weir and Mond [8]. We show that how to build such sets into  $R^n$  using invex sets in a lower dimensional space.

The main results proved in this note relate a differentiable function satisfying Hanson condition to a condition called preinvexity by Jeyakumar [4]. The notion of a preinvex function is defined in the next section, which also presents the main results.

## 2. INVEX SETS

**DEFINITION 2.1.** We call a set  $A \subseteq R^n$  invex with respect to a given  $\eta : R^n \times R^n \rightarrow R^n$  if

$$x, y \in A, 0 \leq \lambda \leq 1 \Rightarrow y + \lambda \eta(x, y) \in A.$$

*Remark 2.1.* It is to be noted that any set in  $R^n$  is invex with respect to  $\eta(x, y) \equiv 0 \forall x \in R^n, y \in R^n$ . However, the only function  $f : R^n \rightarrow R$  invex with respect to  $\eta$  is the trivial function  $f(x) \equiv c \forall x \in R^n$ , where  $c$  is a real number.

The definition essentially says that there is a path starting from  $y$  which is contained in  $A$ . We do not require that  $x$  should be one of the end points

of the path. However, if we demand that  $x$  should be an end point of the path for every pair  $x, y$  then  $\eta(x, y) = x - y$ , reducing to convexity.

**EXAMPLE 2.1.** The following is an example of a bounded invex set in  $R$ , which is invex with respect to a nontrivial  $\eta : R \times R \rightarrow R$ . Let us consider the bounded set  $[-7, -2] \cup [2, 10]$ . This set is a bounded invex set with respect to  $\eta$  given as

$$\eta(x, y) = x - y, x \geq 0, y \geq 0$$

$$\eta(x, y) = x - y, x \leq 0, y \leq 0$$

$$\eta(x, y) = -7 - y, x \geq 0, y \leq 0$$

$$\eta(x, y) = 2 - y, x \leq 0, y \geq 0.$$

Examples of an invex set and an invex function in  $R$  have been given in Weir and Mond [8]. The following proposition enables us to construct invex sets in  $R^n$ , starting from an invex set in  $R$ .

**PROPOSITION 2.1.** Suppose that  $S_1 \subseteq R, S_2 \subseteq R$  such that  $S_1$  is invex with respect to  $\eta_1 : R \times R \rightarrow R$  and  $S_2$  is invex with respect to  $\eta_2 : R \times R \rightarrow R$ . Then  $S_1 \times S_2 \subseteq R^2$  is invex with respect to  $\eta : R^2 \times R^2 \rightarrow R^2$  defined by

$$\eta \begin{pmatrix} x_1, y_1 \\ x_2, y_2 \end{pmatrix} = \begin{pmatrix} \eta_1(x_1, y_1) \\ \eta_2(x_2, y_2) \end{pmatrix}.$$

*Proof.* This is easy to verify.

**EXAMPLE 2.2.** The above proposition shows that the following set in  $R^2$  is invex with respect to  $\eta$ :

Let us consider the invex sets  $S_1 = [-5, -2] \cup [2, 7], S_2 = [-7, -2] \cup [2, 10]$  which are invex with respect to  $\eta_1, \eta_2$ , respectively, where  $\eta_1$  and  $\eta_2$  are given as

$$\eta_1(x, y) = x - y, x \geq 0, y \geq 0 \quad \eta_2(x, y) = x - y, x \geq 0, y \geq 0$$

$$\eta_1(x, y) = x - y, x \leq 0, y \leq 0 \quad \eta_2(x, y) = x - y, x \leq 0, y \leq 0$$

$$\eta_1(x, y) = -5 - y, x \geq 0, y \leq 0 \quad \eta_2(x, y) = -7 - y, x \geq 0, y \geq 0$$

$$\eta_1(x, y) = 2 - y, x \leq 0, y \geq 0 \quad \eta_2(x, y) = 2 - y, x \leq 0, y \geq 0.$$

Clearly,  $S_1 \times S_2$  is invex with respect to  $\eta = \begin{pmatrix} \eta_1 \\ \eta_2 \end{pmatrix}$ .

The following is an example of an invex set in  $R^2$  which is not a cartesian product of two invex sets in  $R$ .

**EXAMPLE 2.3.** Let us consider the set  $\{(u, v) \mid u \geq 0, v \geq 0, u + v \leq 3\} \cup \{(u, v) \mid u \geq 0, v \geq 0, 3u - 2v \geq 9, u \leq 5\}$ . Also let  $\eta = \begin{pmatrix} \eta_1 \\ \eta_2 \end{pmatrix}$ . This set is invex with respect to the function  $\eta$  specified as

$$\begin{aligned} \eta_1 &= x_1 - y_1, & 0 \leq x_1 \leq 3, & 0 \leq y_1 \leq 3 \\ \eta_2 &= x_2 - y_2, & 0 \leq x_2 \leq (3 - x_1), & 0 \leq y_2 \leq (3 - x_1) \\ \eta_1 &= x_1 - y_1, & 3 \leq x_1 \leq (3 + \frac{2}{3}x_2), & 3 \leq y_1 \leq (3 + \frac{2}{3}y_2) \\ \eta_2 &= x_2 - y_2, & 0 \leq x_2 \leq 3, & 0 \leq y_2 \leq 3 \\ \eta_1 &= -y_1, & 3 \leq x_1 \leq (3 + \frac{2}{3}x_2), & 0 \leq y_1 \leq 3 \\ \eta_2 &= -y_2, & 0 \leq x_2 \leq 3, & 0 \leq y_2 \leq (3 - y_1) \\ \eta_1 &= 3 - y_1, & 0 \leq x_1 \leq 3, & 3 \leq y_1 \leq (3 + \frac{2}{3}y_2) \\ \eta_2 &= -y_2, & 0 \leq x_2 \leq (3 - x_1), & 0 \leq y_2 \leq 3. \end{aligned}$$

The general problem of identifying classes of invex sets in  $R^n$  that are useful in the theory of optimization remains open. In what follows we consider nondifferentiable functions which have a convex like property over an invex set. Such functions have been called preinvex by Jeyakumar [4].

**DEFINITION 2.2.** Let  $A \subseteq R^n$  be an invex set, with respect to  $\eta : R^n \times R^n \rightarrow R^n$ . Let  $f : A \rightarrow R$ . We say that  $f$  is preinvex if  $f(x^2 + \lambda\eta(x^1, x^2)) \leq \lambda f(x^1) + (1 - \lambda)f(x^2)$ ,  $\forall x_1, x_2 \in A$ ,  $0 \leq \lambda \leq 1$ .

**DEFINITION 2.3.** Let  $A \subseteq R^n$  be an invex set, with respect to  $\eta : R^n \times R^n \rightarrow R^n$ . We say that  $f$  is prequasiinvex with respect to  $\eta$  if  $x^1, x^2 \in A$ ,  $0 \leq \lambda \leq 1$ , implies that  $f(x^2 + \lambda\eta(x^1, x^2)) \leq \max(f(x^1), f(x^2))$  for all  $x_1, x_2 \in A$ ,  $0 \leq \lambda \leq 1$ .

Also we say that  $f$  is prepseudoinvex with respect to  $\eta$  if  $f(x^1) < f(x^2) \Rightarrow f(x^2 + \lambda\eta(x^1, x^2)) \leq \lambda f(x^1) + \lambda(1 - \lambda)b(x^1, x^2)$  for all  $0 \leq \lambda \leq 1$ , where  $b : R^n \times R^n \rightarrow R^1$  is a positive function. These definitions are due to Pini [7].

Pini [7] shows that if  $f$  is defined on an invex set  $A$  and if it is preinvex and differentiable then  $f$  is also invex with respect to  $\eta$ . The converse is not true in general and Pini [7] gives an example.

In what follows, we shall show that with the following condition imposed on  $\eta$ , a differentiable function which is invex on  $A$ , with respect to  $\eta$ , is also preinvex.

**Condition C.** Let  $\eta : R^n \times R^n \rightarrow R^n$ ; we say that the function  $\eta$  satisfies the condition C if for any  $x^1, x^2$ ,

$$\begin{aligned} \eta(x^2, x^2 + \lambda\eta(x^1, x^2)) &= -\lambda\eta(x^1, x^2), \\ \eta(x^1, x^2 + \lambda\eta(x^1, x^2)) &= (1 - \lambda)\eta(x^1, x^2) \end{aligned}$$

for all  $0 \leq \lambda \leq 1$ .

**THEOREM 2.1.** *Suppose that  $A$  is a preinvex set with respect to  $\eta$  and suppose that  $f : X \rightarrow R$  is differentiable where  $X$  is open and  $X \supseteq A$ . Further suppose that  $f$  is invex with respect to  $\eta$  on  $A$  and that  $\eta$  satisfies condition C. Then  $f$  is preinvex with respect to  $\eta$  on  $A$ .*

*Proof.* Suppose that  $x^1, x^2 \in A$ . Let  $0 < \lambda < 1$  be given and look at  $\bar{x} = x^2 + \lambda\eta(x^1, x^2)$ . Note that  $\bar{x} \in A$ . By, the invexity of  $f$  we have

$$f(x^1) - f(\bar{x}) \geq \eta(x^1, \bar{x})' \nabla f(\bar{x}). \quad (2.1)$$

Similarly, the invexity condition applied to the pair  $x^2, \bar{x}$  yields

$$f(x^2) - f(\bar{x}) \geq \eta(x^2, \bar{x})' \nabla f(\bar{x}). \quad (2.2)$$

Now, multiplying (2.1) by  $\lambda$  and (2.2) by  $(1 - \lambda)$  and adding, we note that  $\lambda f(x^1) + (1 - \lambda)f(x^2) - f(\bar{x}) \geq (\lambda\eta(x^1, \bar{x})' + (1 - \lambda)\eta(x^2, \bar{x})' \nabla f(\bar{x})$ . However, by condition C,  $\lambda\eta(x^1, \bar{x})' + (1 - \lambda)\eta(x^2, \bar{x})' = 0$ . Hence, the conclusion of the theorem follows.

*Remark 2.2.* The above proof is similar to the proof in the convex case.

Similarly, we can prove the following.

**THEOREM 2.2.** *Let  $A \subseteq R^n$  be invex with respect to  $\eta$  and let  $f : X \rightarrow R$  be differentiable on  $X$ , where  $X$  is an open set containing  $A$ . Suppose that  $f$  is quasiinvex with respect to  $\eta$  on  $A$  and that  $\eta$  satisfies condition C. Then  $f$  is prequasiinvex on  $A$ .*

*Proof.* Suppose that  $x^1, x^2 \in A$  and let  $f(x^1) \leq f(x^2)$ . Consider the set

$$\Omega = \{x \mid x = x^2 + \lambda\eta(x^1, x^2), f(x) > f(x^2), 0 \leq \lambda \leq 1\}.$$

In order to show that  $f$  is prequasiinvex, we have to show that  $\Omega = \phi$ . Note that if  $\Omega \neq \phi$  then by, continuity of  $f$ , the set

$$\Omega' = \{x \mid x = x^2 + \lambda\eta(x^1, x^2), f(x) > f(x^2), 0 < \lambda < 1\}$$

is also nonempty. Hence, it is sufficient to show that  $\Omega' = \phi$ , to complete the proof.

Suppose now that  $\bar{x} \in \Omega'$ . We then have  $\bar{x} = x^2 + \bar{\lambda}\eta(x^1, x^2)$ , for some  $0 < \bar{\lambda} < 1$  and  $f(\bar{x}) > f(x^2) \geq f(x^1)$ . By the definition of quasiinvexity it follows, considering the pair  $\bar{x}$  and  $x^1$ , that

$$(\eta(x^1, \bar{x}))' \nabla f(\bar{x}) \leq 0. \quad (2.3)$$

Similarly, considering the pair  $x^2$  and  $\bar{x}$ , it follows that

$$(\eta(x^2, \bar{x}))' \nabla f(\bar{x}) \leq 0. \quad (2.4)$$

Hence by condition C, we have

$$-\bar{\lambda} \eta(x^1, x^2)' \nabla f(\bar{x}) \leq 0 \quad (2.5)$$

and

$$(1 - \bar{\lambda}) \eta(x^1, x^2)' \nabla f(\bar{x}) \leq 0. \quad (2.6)$$

Now (2.5) and (2.6), together with the fact that  $0 < \bar{\lambda} < 1$ , imply that

$$\eta(x^2, x^1)' \nabla f(\bar{x}) = 0. \quad (2.7)$$

Note that (2.7) holds for any  $\bar{x} \in \Omega'$ . Now suppose that  $\Omega' \neq \phi$ . Let  $\bar{x} \in \Omega'$  and let

$$\bar{x} = x^2 + \bar{\lambda} \eta(x^1, x^2).$$

By the continuity of  $f$  we can find  $\lambda^* < \bar{\lambda} < \hat{\lambda} < 1$  such that for all  $\lambda \in (\lambda^*, \hat{\lambda})$ , we have

$$\begin{aligned} f(x^2 + \lambda \eta(x^1, x^2)) &> f(x^2), \\ f(x^2 + \lambda^* \eta(x^1, x^2)) &= f(x^2) \end{aligned}$$

(It is possible that  $\lambda^* = 0$ .) Let  $h(\lambda) = f(x^2 + \lambda \eta(x^1, x^2))$ ; we have  $h(\lambda^*) = f(x^2)$ .

Now, by the mean value theorem applied to the function  $h : [\lambda^*, \hat{\lambda}]$  we have

$$h(\bar{\lambda}) - h(\lambda^*) = \left. \frac{dh}{d\lambda} \right|_{\lambda=\tilde{\lambda}},$$

where  $\tilde{\lambda} \in (\lambda^*, \hat{\lambda})$  or

$$f(x^2 + \bar{\lambda} \eta(x^1, x^2)) - f(x^2) = \eta(x^1, x^2)' \nabla f(x^2 + \tilde{\lambda} \eta(x^1, x^2)).$$

The right-hand side is positive by our hypothesis, but the left-hand side is zero by (2.7), as  $x^2 + \tilde{\lambda} \eta(x^1, x^2) \in \Omega'$ , by construction; hence, we have a contradiction. The proof follows in this case. The proof is similar in case  $f(x^2) \leq f(x^1)$ .

*Remark 2.3.* It is easy to show that a differentiable function which is pre-quasiinvex with respect to  $\eta$ , on a set  $A$  which is invex with respect to  $\eta$ , is also quasiinvex. Our theorem 2.2 is a converse of this under condition C.

**COROLLARY 2.1.** *Suppose that  $g : R^n \rightarrow R$  is quasiinvex with respect to  $\eta$ . Further, suppose that condition C is satisfied by  $\eta$  then  $S = \{x \mid g(x) \leq 0\}$  is also invex with respect to  $\eta$ .*

*Proof.* This is clear from pre-quasiinvexity of  $g$  under the condition of the corollary.

The above corollary is useful in constrained minimization. We can easily show the following result.

**THEOREM 2.3.** *Let  $f : R^n \rightarrow R$  be a differentiable and pre-pseudoinvex function on an invex set  $A$  with respect to a function  $\eta : R^n \times R^n \rightarrow R^n$ . Then  $f$  is quasiinvex with respect to  $\eta$ .*

**THEOREM 2.4.** *If  $f : R^n \rightarrow R$  is pre-pseudoinvex then  $f$  is pre-quasiinvex with respect to the same  $\eta$ .*

**THEOREM 2.5.** *Let  $f : R^n \rightarrow R$  be a pre-pseudoinvex function with respect to  $\eta$ . Also assume that  $\phi : R \rightarrow R$  is a nondecreasing function. Then, the composite function  $\phi \circ f$  is pre-pseudoinvex with respect to  $\eta$ .*

**THEOREM 2.6.** *Let  $f : R^n \rightarrow R$  be a pre-pseudoinvex function on an invex set  $A \subseteq R^n$  with respect to  $\eta : R^n \times R^n \rightarrow R^n$ . Assume that  $\eta(x, y) \neq 0$  whenever  $x \neq y$ . Then every strict local minimizer of the function  $f$  is also a strict global minimizer. The set of points which are strict global minimizers is invex with respect to  $\eta$ .*

*Remark 2.4.* Like pre-pseudoinvexity the above two theorems also hold in the case of pre-quasiinvexity (see Pini [7]).

In the following example we can verify that condition C holds. This shows that condition C may hold for a large class of functions  $\eta$ , rather than just for the trivial case  $\eta(x, y) = x - y$ .

**EXAMPLE 2.4.** Consider Example 2.1 for the bounded set  $[-7, -2] \cup [2, 10]$ . In this set condition C holds with respect to  $\eta$  given as

$$\eta(x, y) = x - y, \quad x \geq 0, y \geq 0$$

$$\eta(x, y) = x - y, \quad x \leq 0, y \leq 0$$

$$\eta(x, y) = -7 - y, \quad x \geq 0, y \leq 0$$

$$\eta(x, y) = 2 - y, \quad x \leq 0, y \geq 0.$$

## REFERENCES

1. A. EEN-ISRAEL AND B. MOND, What is Invexity? *Bull. Austral. Math. Soc. Ser. B* **28** (1986), 1–9.
2. B. D. CRAVEN, Invex functions and constrained local minima, *Bull. Austral. Math. Soc. Ser. B* **24** (1981), 357–366.
3. M. HANSON, On sufficiency of the Kuhn Tucker conditions, *J. Math. Anal. Appl.* **80** (1981), 545–550.
4. V. JEYAKUMAR, Strong and weak invexity in mathematical programming, *Methods Oper. Res.* **55** (1985), 109–125.
5. W. KARUSH, "Minima of Functions of Several Variables with Inequalities as Side Conditions," M.S. thesis, Department of Mathematics, University of Chicago, 1939.
6. H. W. KUHN AND A. W. TUCKER, "Nonlinear Programming; Proceedings, 2nd Berkley Symposium on Mathematical Statistics and Probability" (J. Neyman, Ed.), Univ. of California Press, Berkeley, CA, 1951.
7. R. FINI, Invexity and generalized convexity, *Optimization* **22** (1991), 513–525.
8. T. WEIR AND B. MOND, Preinvex functions in multiple objective optimization, *J. Math. Anal. Appl.* **136** (1988), 29–38.