

η -PSEUDOLINEARITY*

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The notion of η -pseudolinearity is introduced. First, some characterizations of an η -pseudolinear function are obtained. Then characterizations of the solution set of an η -pseudolinear program are derived. The paper generalizes various results on pseudolinear functions and programs.

KEYWORDS: η -pseudoconvex function, η -pseudolinear function, solution set.

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1. Introduction

In response to modeling needs in various disciplines, the classical notion of convexity has been generalized in many ways ([1]). Among others, pseudoconvexity introduced by Mangasarian (1965), proved to be very useful in economic theory and optimization, for example [1]. A real-valued differentiable function f defined on an open set D in \mathbb{R}^n is called pseudolinear ([4]) if f and $-f$ are pseudoconvex. Hanson (1981) considered the class of functions f with the following property:

$$f(y) - f(x) \geq \nabla f(x)^\top \eta(y, x) \quad \text{for all } y, x \in D, \quad (1)$$

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for some given vector-valued function $\eta(y, x)$ defined on $D \times D$. Subsequently, Craven (1981a, 1981b) called the functions satisfying (1) "invex functions" while Kaul and Kaur (1985) called such functions " η -convex functions".

To generalize both η -convexity and pseudoconvexity, Hanson (1981) introduced a more general class of functions defined in the following way:

$$\nabla f(x)^\top \eta(y, x) \geq 0 \quad \text{implies} \quad f(y) \geq f(x) \quad \text{for all } x, y \in D. \quad (2)$$

Later Kaul and Kaur (1985) called functions satisfying (2) " η -pseudoconvex functions" while Craven (1981a) called such functions "pseudoinvex functions". Ben-Israel and Mond (1986) pointed out, the class of pseudoinvex functions coincides with the class of invex functions. But it should be noted that a pseudoinvex function may not be invex with respect to the same vector function η . In order to avoid confusion, we will adopt the notion of η -pseudoconvexity in this paper.

Clearly, f is η -pseudoconvex on D if and only if $f(y) < f(x)$ implies $\nabla f(x)^\top \eta(y, x) < 0$ for all $x, y \in D$.

If $\eta(y, x) = y - x$ for all $x, y \in D$, then the definitions of η -convexity and η -pseudoconvexity reduce to the definitions of convexity and pseudoconvexity, respectively.

There is a sizable literature on pseudolinear functions; see for example [1, 3, 4, 8, 10, 11, 13] and the references therein. In this paper we introduce and study the following generalization.

DEFINITION 1. A differentiable function f defined on an open set D in \mathbb{R}^n is called η -pseudolinear if f and $-f$ are η -pseudoconvex with respect to the same η .

We note that every pseudolinear function is η -pseudolinear with $\eta(x, y) = x - y$, but the converse is not true. The function in the following example is η -pseudolinear but not pseudolinear.

EXAMPLE 1. Let $D = \{(x_1, x_2) \in \mathbb{R} \times \mathbb{R} : x_1 > -1, -\frac{\pi}{2} < x_2 < \frac{\pi}{2}\}$ and $\eta : D \times D \rightarrow \mathbb{R}^2$ defined as follows

$$\eta(y, x) = \left(y_1 - x_1, \frac{\sin y_2 - \sin x_2}{\cos x_2} \right)^\top \quad \text{for all } x = (x_1, x_2), y = (y_1, y_2) \in D.$$

Then the function $f : D \rightarrow \mathbb{R}$ defined by

$$f(x) = x_1 + \sin x_2 \quad \text{for all } x = (x_1, x_2) \in D$$

is η -pseudolinear but not pseudolinear. To see the latter, take $x = (\frac{\pi}{8}, \frac{\pi}{8})$ and $y = (\frac{\pi}{3}, 0)$. Then $\nabla f(x)^\top \eta(y, x) = 0$, but $f(y) < f(x)$.

DEFINITION 2. [14] For a given $\eta : K \times K \rightarrow \mathbb{R}^n$, a nonempty subset K of \mathbb{R}^n is called η -convex (or invex), if for each $x, y \in K$, $0 \leq t \leq 1$, $x + t\eta(y, x) \in K$.

DEFINITION 3. [15] Let $\eta : K \times K \rightarrow \mathbb{R}^n$ be a given function and K be a nonempty η -convex subset of \mathbb{R}^n . A function $f : K \rightarrow \mathbb{R}$ is said to be pre-invex on K if

$$f(x + t\eta(y, x)) \leq tf(y) + (1 - t)f(x) \quad \text{for all } t \in [0, 1].$$

Many results in this paper assume that the function $\eta : K \times K \rightarrow \mathbb{R}^n$ satisfies condition C in [14], i.e., for any $x, y \in K$

$$\eta(x, x + t\eta(y, x)) = -t\eta(y, x),$$

$$\eta(y, x + t\eta(y, x)) = (1 - t)\eta(y, x)$$

for all $t \in [0, 1]$.

This condition ensures that an η -convex (invex) function is also pre-invex ([14]). For an example of a function η which satisfies condition C, see [14, Example 2.4].

2. Characterizations of η -Pseudolinear Functions

In this section, we provide some characterizations of η -pseudolinear functions.

PROPOSITION 1. *Let f be a differentiable function defined on an open set D in \mathbb{R}^n and K be an η -convex subset of D such that $\eta : K \times K \rightarrow \mathbb{R}^n$ satisfies condition C. Suppose that f is η -pseudolinear on K . Then for all $x, y \in K$, $\nabla f(x)^\top \eta(y, x) = 0$ if and only if $f(x) = f(y)$.*

Proof. Suppose that f is η -pseudolinear on K . Then for all $x, y \in K$, we have

$$\nabla f(x)^\top \eta(y, x) \geq 0 \quad \text{implies} \quad f(y) \geq f(x)$$

and

$$\nabla f(x)^\top \eta(y, x) \leq 0 \quad \text{implies} \quad f(y) \leq f(x).$$

Combining these two inequalities, we obtain

$$\nabla f(x)^\top \eta(y, x) = 0 \quad \text{implies} \quad f(x) = f(y) \quad \text{for all } x, y \in K.$$

Now we prove that $f(x) = f(y)$ implies $\nabla f(x)^\top \eta(y, x) = 0$ for all $x, y \in K$. For that, we show that for any $x, y \in K$ such that $f(x) = f(y)$ implies that $f(x + t\eta(y, x)) = f(x)$ for all $t \in (0, 1)$.

If $f(x + t\eta(y, x)) > f(x)$, then by the definition of η -pseudoconvexity of f we have

$$\nabla f(z)^\top \eta(x, z) < 0 \tag{3}$$

where $z = x + t\eta(y, x)$.

We show that $\eta(x, z) = \frac{-t}{1-t}\eta(y, z)$. From condition C, we have

$$\begin{aligned} \eta(x, z) &= \eta(x, x + t\eta(y, x)) = -t\eta(y, x) \\ &= \frac{-t}{1-t}\eta(y, z). \end{aligned}$$

Therefore from (3), we obtain

$$\nabla f(z)^\top \left(\frac{-t}{1-t} \right) \eta(y, z) < 0$$

and hence $\nabla f(z)^\top \eta(y, z) > 0$. By η -pseudoconvexity of f , we have

$$f(y) \geq f(z).$$

This contradicts the assumption that

$$f(z) > f(x) = f(y).$$

Similarly, we can also show that $f(x + t\eta(y, x)) < f(x)$ leads to a contradiction, using η -pseudoconvexity of $-f$. This proves the claim that $f(x + t\eta(y, x)) = f(x)$ for all $t \in (0, 1)$. Thus

$$\nabla f(x)^\top \eta(y, x) = \lim_{t \rightarrow 0^+} \frac{f(x + t\eta(y, x)) - f(x)}{t} = 0. \quad \blacksquare$$

Now we give an example where the converse of Proposition 1 is not true, that is, if for all $x, y \in K$, $\nabla f(x)^\top \eta(y, x) = 0$ if and only if $f(x) = f(y)$, then f need not be η -pseudolinear.

EXAMPLE 2. Let $D = K = (-\infty, +\infty)$ and $f : D \rightarrow \mathbb{R}$, $\eta : D \times D \rightarrow \mathbb{R}$ be defined as follows

$$f(x) = e^x, \quad \eta(y, x) = e^{-y} - e^{-x}.$$

Then $\nabla f(x)^\top \eta(y, x) = 0 \Leftrightarrow y = x \Leftrightarrow f(x) = f(y)$. But for $x = 2$ and $y = 1$, we have

$$\nabla f(x)^\top \eta(y, x) = e^2(e^{-1} - e^{-2}) = e - 1 > 0$$

and $f(y) = e < e^2 = f(x)$. Hence f is not η -pseudoconvex on D .

PROPOSITION 2. Let f be a differentiable function defined on an open set D in \mathbb{R}^n and K be an η -convex subset of D . Then f is η -pseudolinear on K if and only if there exists a function p defined on $K \times K$ such that $p(x, y) > 0$ and $f(y) = f(x) + p(x, y)\nabla f(x)^\top \eta(y, x)$ for all $x, y \in K$.

Proof. Let f be an η -pseudolinear function. We have to construct a function p on $K \times K$ such that $p(x, y) > 0$ and $f(y) = f(x) + p(x, y)\nabla f(x)^\top \eta(y, x)$ for all $x, y \in K$.

If $\nabla f(x)^\top \eta(y, x) = 0$ for $x, y \in K$, then we define $p(x, y) = 1$. In this case we have $f(y) = f(x)$, due to Proposition 1. On the other hand, if $\nabla f(x)^\top \eta(y, x) \neq 0$, then we define

$$p(x, y) = \frac{f(y) - f(x)}{\nabla f(x)^\top \eta(y, x)}.$$

We have to show that $p(x, y) > 0$. Suppose that $f(y) > f(x)$. Then by η -pseudoconvexity of $-f$, we have $\nabla f(x)^\top \eta(y, x) > 0$. Hence $p(x, y) > 0$. Similarly, if $f(y) < f(x)$, then we have $\nabla f(x)^\top \eta(y, x) < 0$ by η -pseudoconvexity of f . Therefore $p(x, y) > 0$.

To prove the converse, we first show that f is η -pseudoconvex, i.e., for any $x, y \in K$,

$$\nabla f(x)^\top \eta(y, x) \geq 0 \text{ implies } f(y) \geq f(x).$$

If $\nabla f(x)^\top \eta(y, x) \geq 0$, then we have

$$f(y) - f(x) = p(x, y) \nabla f(x)^\top \eta(y, x) \geq 0.$$

Thus $f(y) \geq f(x)$. Likewise, we can prove that $-f$ is η -pseudoconvex. Hence f is η -pseudolinear. ■

REMARK 1. Proposition 1 generalizes an early result by Kortanek and Evans (1967), see also [4]. Proposition 2 generalizes a result by Chew and Choo (1984).

PROPOSITION 3. Let $f : D \rightarrow \mathbb{R}^n$ be an η -pseudolinear function defined on an open set D of \mathbb{R}^n and let $F : \mathbb{R} \rightarrow \mathbb{R}$ be differentiable with $F'(t) > 0$ on $F'(t) < 0$ for all $t \in \mathbb{R}$. Then the composite function $F \circ f$ is also η -pseudolinear.

Proof. Let $g(x) = F(f(x))$ for all $x \in D$. It suffices to prove the result for $F'(t) > 0$ since the negative of an η -pseudolinear function is η -pseudolinear. We have

$$\nabla g(x)^\top \eta(y, x) = F'(f(x)) \nabla f(x)^\top \eta(y, x).$$

Then $\nabla g(x)^\top \eta(y, x) \geq 0$ (≤ 0) implies $\nabla f(x)^\top \eta(y, x) \geq 0$ (≤ 0) since F is strictly increasing. This yields $f(y) \geq f(x)$ ($f(y) \leq f(x)$), due to η -pseudolinearity of f . Thus $g(y) \geq g(x)$ ($g(y) \leq g(x)$) since F is strictly increasing. Hence g is η -pseudolinear. ■

The following example shows that Proposition 3 no longer holds if $F'(t) = 0$ for some t .

EXAMPLE 3. Let f, η and D be defined as in Example 1 and let $F(t) = t^3$ defined on \mathbb{R} . Obviously, $F'(0) = 0$.

For $x = (0, 0)$, $y = (0, -\frac{\pi}{3})$, we have $\nabla g(0, 0) = 0$ and therefore, $\nabla g(0, 0)^\top \eta(y, x) = 0$. But $g(y) = F(f(y)) = (\sin(-\frac{\pi}{3}))^3 = -\frac{3\sqrt{3}}{8} < 0 = g(x)$. Thus g is not η -pseudoconvex, so not η -pseudolinear.

3. Characterizations of Solution Sets

We consider the following problem:

$$\min f(x) \quad \text{subject to } x \in K \quad (\text{P})$$

where $f : D \rightarrow \mathbb{R}$, D is an open subset of \mathbb{R}^n , and K is an η -convex set of D .

We assume throughout this section that the solution set

$$\tilde{S} := \arg \min_{x \in K} f(x)$$

is nonempty.

PROPOSITION 4. If f is a pre-invex function on K , then the solution set \tilde{S} of problem (P) is an η -convex set.

Proof. Let $x_1, x_2 \in \bar{S}$. Then $f(x_1) \leq f(y)$ and $f(x_2) \leq f(y)$ for all $y \in K$. Since f is pre-invex, we have

$$\begin{aligned} f(x_1 + t\eta(x_2, x_1)) &\leq tf(x_2) + (1-t)f(x_1), \quad \text{for all } t \in [0, 1] \\ &\leq tf(y) + (1-t)f(y) \\ &= f(y). \end{aligned}$$

Hence $x_1 + t\eta(x_2, x_1) \in \bar{S}$, and so, \bar{S} is an η -convex set. \blacksquare

REMARK 2. From the proof of Proposition 1, it is easy to show that the solution set \bar{S} of problem (P) is η -convex if $f : D \rightarrow \mathbb{R}$ is η -pseudolinear where $\eta : K \times K \rightarrow \mathbb{R}^n$ satisfies condition C.

Now we state a first-order characterization of the solution set of an η -pseudolinear program in terms of any of its solutions.

THEOREM 1. Let $f : D \rightarrow \mathbb{R}$ be differentiable on an open set D and let f be η -pseudolinear on an η -convex subset $K \subset D$ where η satisfies condition C and $\eta(x, y) + \eta(y, x) = 0$ for all $x, y \in K$. Let $\bar{x} \in \bar{S}$. Then $\bar{S} = \tilde{S} = \hat{S}$ where

$$\tilde{S} := \{x \in K : \nabla f(x)^\top \eta(\bar{x}, x) = 0\}, \quad (4)$$

$$\hat{S} := \{x \in K : \nabla f(\bar{x})^\top \eta(\bar{x}, x) = 0\}. \quad (5)$$

Proof. The point $x \in \bar{S}$ if and only if $f(x) = f(\bar{x})$. By Proposition 1, we have $f(x) = f(\bar{x})$ if and only if $\nabla f(x)^\top \eta(\bar{x}, x) = 0$. Also $f(\bar{x}) = f(x)$ if and only if $\nabla f(\bar{x})^\top \eta(x, \bar{x}) = 0$. The latter is equivalent to $\nabla f(\bar{x})^\top \eta(\bar{x}, x) = 0$ since $\eta(\bar{x}, x) = -\eta(x, \bar{x})$. \blacksquare

COROLLARY 1. Let f and η be the same as in Theorem 1. Then $\bar{S} = \tilde{S}_1 = \hat{S}_1$ where

$$\tilde{S}_1 := \{x \in K : \nabla f(x)^\top \eta(\bar{x}, x) \geq 0\},$$

$$\hat{S}_1 := \{x \in K : \nabla f(\bar{x})^\top \eta(\bar{x}, x) \geq 0\}.$$

Proof. It is clear from Theorem 1 that $\bar{S} \subset \tilde{S}_1$. We prove that $\tilde{S}_1 \subset \bar{S}$. Assume that $x \in \tilde{S}_1$, that is,

$$x \in K \quad \text{such that} \quad \nabla f(x)^\top \eta(\bar{x}, x) \geq 0.$$

In view of Proposition 2, there exists a function p defined on $K \times K$ such that $p(x, \bar{x}) > 0$ and

$$f(\bar{x}) = f(x) + p(x, \bar{x}) \nabla f(x)^\top \eta(\bar{x}, x) \geq f(x).$$

This implies that $x \in \bar{S}$, and hence $\tilde{S}_1 \subset \bar{S}$. Similarly we can prove that $\bar{S} = \hat{S}_1$, using the identity $\eta(x, \bar{x}) = -\eta(\bar{x}, x)$. \blacksquare

THEOREM 2. In problem (P), assume that f is differentiable on D and η -pseudolinear on an η -convex set $K \subset D$ where η satisfies condition C and $\eta(x, y) + \eta(y, x) = 0$ for all $x, y \in K$. If $\bar{x} \in \bar{S}$, then $\bar{S} = S^* = S_1^*$ where

$$S^* := \{x \in K : \nabla f(\bar{x})^\top \eta(\bar{x}, x) = \nabla f(x)^\top \eta(x, \bar{x})\},$$

$$S_1^* := \{x \in K : \nabla f(\bar{x})^\top \eta(\bar{x}, x) \geq \nabla f(x)^\top \eta(x, \bar{x})\}.$$

Proof. (i) $\bar{S} \subset S^*$. Let $x \in \bar{S}$. It follows from Theorem 1 that

$$\nabla f(x)^\top \eta(\bar{x}, x) = 0 = \nabla f(\bar{x})^\top \eta(\bar{x}, x).$$

Since $\eta(\bar{x}, x) = -\eta(x, \bar{x})$, we have

$$\nabla f(x)^\top \eta(x, \bar{x}) = 0 = \nabla f(\bar{x})^\top \eta(\bar{x}, x).$$

Thus $x \in S^*$, and hence $\bar{S} \subset S^*$.

(ii) $S^* \subset S_1^*$ is obvious.

(iii) $S_1^* \subset \bar{S}$. Assume that $x \in S_1^*$. Then $x \in K$ satisfies

$$\nabla f(\bar{x})^\top \eta(\bar{x}, x) \geq \nabla f(x)^\top \eta(x, \bar{x}). \quad (6)$$

Suppose that $x \notin \bar{S}$. Then $f(x) > f(\bar{x})$. By η -pseudoconvexity of $-f$, we have

$$\nabla f(\bar{x})^\top \eta(x, \bar{x}) > 0.$$

Since $\eta(x, \bar{x}) = -\eta(\bar{x}, x)$, we have $\nabla f(\bar{x})^\top \eta(\bar{x}, x) < 0$. Using (6), we have

$$\nabla f(x)^\top \eta(x, \bar{x}) < 0 \quad \text{or} \quad \nabla f(x)^\top \eta(\bar{x}, x) > 0.$$

In view of Proposition 2, there exists a function p defined on $K \times K$ such that $p(x, \bar{x}) > 0$, and

$$f(\bar{x}) = f(x) + p(x, \bar{x}) \nabla f(x)^\top \eta(\bar{x}, x) > f(x),$$

a contradiction. Hence $x \in \bar{S}$. ■

REMARK 3. This section generalizes results by Jeyakumar and Yang (1995).

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RIASSUNTO

Il lavoro introduce la nozione di η -pseudolinearità. Dopo avere ottenuto alcune caratterizzazioni delle funzioni η -pseudolineari, si derivano caratterizzazioni dell'insieme delle soluzioni di un programma η -pseudolineare. Lo studio generalizza diversi risultati sulle funzioni e sui programmi pseudolineari.

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