



## SUFFICIENT OPTIMALITY CONDITIONS IN PROBLEMS WITH $(h, \varphi) - (p, r)$ -INVEX FUNCTIONS

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The notion of  $p$ -invex sets with respect to  $\eta$  is extended for generalized algebraic operations of Ben-Tal to  $h-p$ -invex sets. A class of real valued functions, called  $(h, \varphi) - (p, r)$ -pre-invex functions (without differentiability) or  $(h, \varphi) - (p, r)$ -invex functions with respect to  $\eta$  (in the differentiable case) is introduced. Sufficient optimality conditions are obtained for a nonlinear programming problem involving  $(h, \varphi) - (p, r)$ -invex functions with respect to  $\eta$ .

*Key words:*  $h-p$ -invex set with respect to  $\eta$ ,  $(h, \varphi) - (p, r)$ -pre-invex functions with respect to  $\eta$ ,  $(h, \varphi) - (p, r)$ -invex functions with respect to  $\eta$ .

### 1. INTRODUCTION

In this article we introduce  $h-p$ -invex sets with respect to a function  $\eta: X \times X \rightarrow R^n$ . Based on this definition, we define new classes of (nonconvex) functions which we call  $(h, \varphi) - (p, r)$ -pre-invex functions with respect to  $\eta$ , and their equivalents in the case of  $(h, \varphi)$ -differentiability –  $(h, \varphi) - (p, r)$ -invex functions with respect to  $\eta$ . The  $(h, \varphi) - (p, r)$ -pre-invex functions are a generalization of the  $(p, r)$ -pre-invex functions if we take  $h(x) \equiv x$ , and  $\varphi(\alpha) \equiv \alpha$ , which, are a generalization of pre-invex functions with respect to  $\eta$  introduced by Ben-Israel and Mond [2], whereas the  $(h, \varphi) - (p, r)$ -invex functions are a generalization of  $(p, r)$ -invex functions, which contain invex functions with respect to  $\eta$  introduced by Hanson [3].

The next part of considerations is developed for optimality conditions in nondifferentiable and differentiable mathematical programming problems. The sufficient optimality conditions are obtained in  $(h, \varphi)$ -mathematical programming problems with inequality constraints in which the functions occurring belong to the class of functions introduced in this article. The results obtained here extend those well-known so far; see Hanson [3].

### 2. PRELIMINARIES

**DEFINITION 2.1.** Let  $\alpha \in R^m$ ,  $q \in R^m$  be vectors, whose coordinates are positive and nonnegative numbers, respectively, and let  $r$  be an arbitrary real number. If we assume that

$$\sum_{i=1}^m q_i = 1,$$

then the weighted  $r$ -mean is defined as

$$M_r(\alpha; q) = M_r(\alpha_1, \dots, \alpha_m; q) = \begin{cases} \left( \sum_{i=1}^m q_i \alpha_i^r \right)^{\frac{1}{r}} & \forall r \neq 0 \\ \prod_{i=1}^m \alpha_i^{q_i} & r = 0. \end{cases}$$

**Definition 2.2.** We say  $S \subset R^n$  is a  $h-p$ -invex set with respect to a vector function  $\eta: S \times S \rightarrow R^n$  if, for any  $x, u \in S$  and  $q_1 \geq 0, q_2 \geq 0, q_1 + q_2 = 1$ , we have

$$\log \left( M_p \left( e^{\eta_1(x,u)+h_1(u)}, e^{h_1(u)}; q \right), \dots, M_p \left( e^{\eta_n(x,u)+h_n(u)}, e^{h_n(u)}; q \right) \right) \in S.$$

Here the logarithm is understood to be taken componentwise. If  $q_1 = \lambda$  (the fact that  $q_1 + q_2 = 1$  implies that  $q_2 = 1 - \lambda$  for any  $\lambda \in [0; 1]$ ) then, using the definition of weighted  $p$ -mean, we may write down the above relations as

$$\begin{aligned} \log \left( \lambda e^{p\eta(x,u)+ph(u)} + (1-\lambda) e^{ph(u)} \right)^{\frac{1}{p}} &\in S, & p \neq 0, \\ h(u) + \lambda \eta(x, u) &\in S, & p = 0. \end{aligned}$$

Again, the logarithm and exponentials are understood to be taken componentwise.

**Definition 2.3.** Let  $\eta: S \times S \rightarrow R^n$  be a vector function. A function  $f: S \rightarrow R$  defined on a  $h-p$ -invex set  $S \subset R^n$  with respect to  $\eta$  is called  $(h, \varphi) - (p, r)$ -pre-invex with respect to  $\eta$  at the point  $u \in S$  on  $S$  if for any  $x \in S$ ,  $q_1 \geq 0, q_2 \geq 0, q_1 + q_2 = 1$ , the following inequality is satisfied

$$\hat{f} \left( \log M_p \left( e^{\eta(x,u)+h(u)}, e^{h(u)}; q \right) \right) \leq \log M_r \left( e^{\varphi(f(x))}, e^{\varphi(f(u))}; q \right),$$

where  $\hat{f}(t) = \varphi(f(h^{-1}(t)))$  is the function appearing in the definition of the  $(h, \varphi)$ -differential. If the mentioned inequality is satisfied at any point  $u \in S$ , then  $f$  is said to be  $(h, \varphi) - (p, r)$ -pre-invex with respect to  $\eta$  on  $S$ .

If  $q_1 = \lambda$  (the fact that  $q_1 + q_2 = 1$  implies that  $q_2 = 1 - \lambda$  for any  $\lambda \in [0; 1]$ ) then, using the definition of weighted  $p$ -mean, we may write down inequality above as

$$\begin{aligned} \hat{f} \left( \log \left( \lambda e^{p(\eta(x,u)+h(u))} + (1-\lambda) e^{ph(u)} \right)^{\frac{1}{p}} \right) &\leq \log \left( \lambda e^{r\varphi(f(x))} + (1-\lambda) e^{r\varphi(f(u))} \right)^{\frac{1}{r}}, & p \neq 0, r \neq 0, \\ \hat{f} \left( \log \left( \lambda e^{p(\eta(x,u)+h(u))} + (1-\lambda) e^{ph(u)} \right)^{\frac{1}{p}} \right) &\leq \varphi \left( \lambda [\cdot] f(x) [+](1-\lambda) [\cdot] f(u) \right), & p \neq 0, r = 0, \\ \hat{f} \left( \lambda \eta(x, u) + h(u) \right) &\leq \log \left( \lambda e^{r\varphi(f(x))} + (1-\lambda) e^{r\varphi(f(u))} \right)^{\frac{1}{r}}, & p = 0, r \neq 0, \\ \hat{f} \left( \lambda \eta(x, u) + h(u) \right) &\leq \varphi \left( \lambda [\cdot] f(x) [+](1-\lambda) [\cdot] f(u) \right), & p = 0, r = 0. \end{aligned} \tag{1}$$

*Remark.* All classes of functions defined by (1) above according to **Definition 2.3** are called  $(h, \varphi) - (p, r)$ -pre-invex with respect to  $\eta$ . But one may use the terminology below.

- In case  $p \neq 0, r = 0$ , functions defined by (1) are called  $(h, \varphi) - (p, 0)$ -pre-invex with respect to  $\eta$  (or shortly  $(h, \varphi) - p$ -pre-invex with respect to  $\eta$ );
- In case  $p = 0, r \neq 0$ , functions defined by (1) are called  $(h, \varphi) - (0, r)$ -pre-invex with respect to  $\eta$  (or shortly  $(h, \varphi) - r$ -pre-invex with respect to  $\eta$ );

- In case  $p = 0, r = 0$ , functions defined by (1) are called  $(h, \varphi) - (0, 0)$ -pre-invex with respect to  $\eta$  (or shortly  $(h, \varphi)$ -pre-invex with respect to  $\eta$ );

**Definition 2.4.** Let  $f : S \rightarrow R$  be a differentiable function on an  $h - p$ -invex set  $S \subset R^n$  with respect to  $\eta$ . If for all  $x \in S$  one of the following relations holds:

$$\begin{aligned} \frac{1}{r} e^{r\varphi(f(x))} &\geq \frac{1}{r} e^{r\varphi(f(u))} \left( 1 + \frac{r}{p} h(\nabla^* f(u)) (e^{p\eta(x,u)} - 1) \right), & p \neq 0, \quad r \neq 0, \\ \varphi(f(x)[-]f(u)) &\geq \frac{1}{p} h(\nabla^* f(u)) (e^{p\eta(x,u)} - 1), & p \neq 0, \quad r = 0, \\ \frac{1}{r} e^{r\varphi(f(x))} &\geq \frac{1}{r} e^{r\varphi(f(u))} (1 + rh(\nabla^* f(u))\eta(x,u)), & p = 0, \quad r \neq 0, \\ \varphi(f(x)[-]f(u)) &\geq h(\nabla^* f(u))\eta(x,u), & p = 0, \quad r = 0. \end{aligned} \quad (2)$$

then  $f$  is said to be  $(h, \varphi) - (p, r)$ -invex with respect to  $\eta$  at  $u$  on  $S$ .

**THEOREM 2.1.** Let  $S \subset R^n$  be a  $h - p$ -invex set with respect to  $\eta$ , and let  $f : S \rightarrow R$  be a differentiable function. If  $f$  is  $(h, \varphi) - (p, r)$ -pre-invex with respect to  $\eta$  on  $S$ , then  $f$  is  $(h, \varphi) - (p, r)$ -invex with respect to  $\eta$  on  $S$ .

*Proof.* Let  $f : S \rightarrow R$  be defined on a  $h - p$ -invex set  $S \subset R^n$  with respect to  $\eta$ . Moreover, we assume that  $f$  is a  $(h, \varphi)$ -differentiable  $(h, \varphi) - (p, r)$ -pre-invex function with respect to  $\eta$  on  $S$ . There are four possible cases.

- **Case 1:**  $p \neq 0, r \neq 0$ ,
- **Case 2:**  $p \neq 0, r = 0$ ,
- **Case 3:**  $p = 0, r \neq 0$ ,
- **Case 4:**  $p = 0, r = 0$ ,

We give the proof for each case.

- **Case 1:  $p \neq 0, r \neq 0$ .** Assume  $r > 0$ . By Definition 2.3 of  $(h, \varphi) - (p, r)$ -pre-invex functions we have

$$\begin{aligned} e^{\hat{f} \left( \log \left( \lambda e^{p(\eta(x,u)+h(u))} + (1-\lambda)e^{ph(u)} \right)^{\frac{1}{p}} \right)} &\leq \left( \lambda e^{r\varphi(f(x))} + (1-\lambda)e^{r\varphi(f(u))} \right)^{\frac{1}{r}} \Rightarrow \\ \Rightarrow e^{\hat{r}\hat{f} \left( \log \left( \lambda e^{p(\eta(x,u)+h(u))} + (1-\lambda)e^{ph(u)} \right)^{\frac{1}{p}} \right)} &\leq \lambda e^{r\varphi(f(x))} + (1-\lambda)e^{r\varphi(f(u))} \Rightarrow \\ \Rightarrow \frac{e^{\hat{r}\hat{f} \left( \log \left( \lambda e^{p(\eta(x,u)+h(u))} + (1-\lambda)e^{ph(u)} \right)^{\frac{1}{p}} \right)} - e^{\hat{r}\hat{f}(h(u))}}{\lambda} &\leq e^{r\varphi(f(x))} - e^{r\varphi(f(u))}. \end{aligned} \quad (3)$$

For simplicity, we define

$$\mu(\lambda) = \log \left( \lambda e^{p(\eta(x,u)+h(u))} + (1-\lambda)e^{ph(u)} \right)^{\frac{1}{p}},$$

We now can easily check that  $\mu(0) = h(u)$ , and by letting  $\lambda \rightarrow 0$  we get

$$e^{r\varphi(f(x))} - e^{r\varphi(f(u))} \geq \lim_{\lambda \rightarrow 0} \frac{e^{r\hat{f}(\mu(\lambda))} - e^{r\hat{f}(\mu(0))}}{\lambda} = \frac{r}{p} e^{r\varphi(f(u))} h(\nabla^* f(u)) (e^{p\eta(x,u)} - 1),$$

which, after algebraic transformations, becomes

$$\frac{1}{r} e^{r\varphi(f(x))} \geq \frac{1}{r} e^{r\varphi(f(u))} \left( 1 + \frac{r}{p} h(\nabla^* f(u)) (e^{p\eta(x,u)} - 1) \right), \quad (4)$$

Thus, according to Definition 2.4,  $f$  is  $(h, \varphi) - (p, r)$ -invex function with respect to  $\eta$  on  $S$  as it satisfies the first condition in (2). The proof of the case when  $r < 0$  is totally analogous, only the inequality direction changes at (3) and then changes back at (4).

- **Case 2:  $p \neq 0, r = 0$ .** By Definition 2.3 of  $(h, \varphi) - (p, r)$ -pre-invex functions we have

$$\begin{aligned} \hat{f} \left( \log \left( \lambda e^{p(\eta(x,u)+h(u))} + (1-\lambda) e^{ph(u)} \right)^{\frac{1}{p}} \right) &\leq \varphi(\lambda [\cdot] f(x) [+](1-\lambda) [\cdot] f(u)) \Rightarrow \\ \frac{\hat{f} \left( \log \left( \lambda e^{p(\eta(x,u)+h(u))} + (1-\lambda) e^{ph(u)} \right)^{\frac{1}{p}} \right) - \hat{f}(h(u))}{\lambda} &\leq \varphi(f(x)) + \varphi(f(u)) \end{aligned}$$

and letting  $\lambda \rightarrow 0$  we get

$$\varphi(f(x) [-] f(u)) \geq \frac{1}{p} h(\nabla^* f(u)) (e^{p\eta(x,u)} - 1),$$

- **Case 3:  $p = 0, r \neq 0$ .** By Definition 2.3 of  $(h, \varphi) - (p, r)$ -pre-invex functions we have

$$\hat{f}(\lambda\eta(x,u) + h(u)) \leq \log \left( \lambda e^{r\varphi(f(x))} + (1-\lambda) e^{r\varphi(f(u))} \right)^{\frac{1}{r}}.$$

Assuming  $r > 0$  we can transform the inequality as

$$\frac{e^{\hat{f}(\lambda\eta(x,u)+h(u))} - e^{\hat{f}(h(u))}}{\lambda} \leq e^{r\varphi(f(x))} - e^{r\varphi(f(u))}, \quad (5)$$

By letting  $\lambda \rightarrow 0$  we get

$$e^{r\varphi(f(x))} - e^{r\varphi(f(u))} \geq r e^{r\varphi(f(u))} (1 + rh(\nabla^* f(u))\eta(x,u)),$$

which after algebraic transformation can be written as

$$\frac{1}{r} e^{r\varphi(f(x))} \geq \frac{1}{r} e^{r\varphi(f(u))} (1 + rh(\nabla^* f(u))\eta(x,u)), \quad (6)$$

Again, similarly to Case 1, the proof for the  $r < 0$  is the same, except that the inequality changes to opposite direction at (5), and then changes back at (6).

- **Case 4:  $p = 0, r = 0$ .** By Definition 2.3 of  $(h, \varphi) - (p, r)$ -pre-invex functions we have

$$\begin{aligned} \hat{f}(\lambda\eta(x,u) + h(u)) &\leq \varphi(\lambda [\cdot] f(x) [+](1-\lambda) [\cdot] f(u)) \Rightarrow \\ \Rightarrow \hat{f}(\lambda\eta(x,u) + h(u)) &\leq \varphi(\lambda [\cdot] f(x)) + \varphi((1-\lambda) [\cdot] f(u)) = \lambda\varphi(f(x)) + (1-\lambda)\varphi(f(u)), \end{aligned}$$

which, after some algebraic transformation can be written as

$$\frac{\hat{f}(\lambda\eta(x,u) + h(u)) - \hat{f}(h(u))}{\lambda} = \varphi(f(x)) - \varphi(f(u)).$$

By letting  $\lambda \rightarrow 0$  we get

$$\begin{aligned} h(\nabla^* f(u))\eta(x, u) &\leq \varphi(f(x)) - \varphi(f(u)) = \varphi(f(x)[-]f(u)) \Rightarrow \\ &\Rightarrow h(\nabla^* f(u))\eta(x, u) \leq \varphi(f(x)[-]f(u)). \end{aligned}$$

### 3. OPTIMALITY CONDITIONS

In this section we shall deal with sufficient optimality conditions in problems with inequality constraints. As it is known, the Karush-Kuhn-Tucker conditions are necessary for optimality in mathematical programming problem, if a certain constraint holds. It is also well-known that if the objective function and the functions of constraints occurring in the optimization problem are convex, the Karush-Kuhn-Tucker conditions become sufficient for optimality [1, 4]. This fact also takes place in the case of wider classes of functions, namely, when the functions occurring in optimization problem are invex with respect to the same  $\eta$  [3].

We shall demonstrate below the similar result for  $(h, \varphi)$ -generalized case, when the occurring functions are a certain kind of  $(h, \varphi)$ - $(p, r)$ -invex with respect to the same  $\eta$ . Below we consider that the functions  $h$  and  $\varphi$  satisfy the conditions  $h(0)=0, \varphi(0)=0$  and that  $\varphi$  is a monotone nondecreasing function.

Consider an optimization problem with inequality constraints of the form

$$\begin{aligned} f(x) &\rightarrow \min \\ g_i(x) &\leq 0, \quad i=1, \dots, m, \end{aligned} \quad (\text{P})$$

where  $f, g_i : X_0 \rightarrow R, i=1, 2, \dots, m$  are  $(h, \varphi)$ -differentiable functions on an open nonempty set  $X_0 \subset R^n$ .

Let us denote by  $D$  the set of the feasible points of (P) i.e., the set of the form

$$D = \{x \in X_0 : g_i(x) \leq 0, \quad i=1, 2, \dots, m\}.$$

**Definition 3.1.** A point  $z \in R^n$  is called feasible for problem (P) if  $z \in X_0$  and  $g_i(z) \leq 0, \quad i=1, \dots, m$ .

Now we redefine the Karush-Kuhn-Tucker conditions [1, 4] for the  $(h, \varphi)$ -case as follows.

**Definition 3.2.** We say that  $(h, \varphi)$ -Karush-Kuhn-Tucker conditions holds at a feasible point  $\bar{x}$  for problem (P), if there exists  $\bar{\xi}_i, i=1, 2, \dots, m$  such that

$$\nabla^* f(\bar{x}) \oplus \bigoplus_{i=1}^m \bar{\xi}_i \otimes \nabla^* g_i(\bar{x}) = 0, \quad (7)$$

$$\left[ \sum_{i=1}^m \right] (\bar{\xi}_i [\cdot] g_i(\bar{x})) = 0, \quad (8)$$

$$\bar{\xi}_i \geq 0, \quad i=1, 2, \dots, m. \quad (9)$$

**THEOREM 3.1.** Assume that the point  $\bar{x} \in X_0$  is feasible for problem (P) and let  $(h, \varphi)$ -Karush-Kuhn-Tucker conditions be satisfied at the point  $(\bar{x}, \bar{\xi})$  if the objective function  $f$  and the constraint functions  $g_i, i=1, 2, \dots, m$  are  $(h, \varphi)$ -invex with respect to  $\eta$  at  $x$  on  $D$ , then  $\bar{x}$  is a global minimum point in problem (P).

*Proof.* Assume that  $x$  is an arbitrary feasible point for problem (P). By assumption,  $f$  and  $g_i$  are  $(h, \varphi)$ - $(p, r)$ -invex with respect to the same function  $\eta$  at  $\bar{x}$  on  $D$ . Therefore, for all  $x, \bar{x} \in D$  the inequalities

$$\frac{1}{r} e^{r\varphi(f(x))} \geq \frac{1}{r} e^{r\varphi(f(\bar{x}))} \left( 1 + \frac{r}{p} h(\nabla^* f(\bar{x})) (e^{p\eta(x, \bar{x})} - 1) \right), \quad (10)$$

$$\frac{1}{r} e^{r\varphi(g_i(x))} \geq \frac{1}{r} e^{r\varphi(g_i(\bar{x}))} \left( 1 + \frac{r}{p} h(\nabla^* g_i(\bar{x})) (e^{p\eta(x, \bar{x})} - 1) \right), \quad (11)$$

are true. Denote  $J = \{1 \leq i \leq m; \bar{\xi}_i > 0\}$ , where  $\bar{\xi} = (\bar{\xi}_1, \bar{\xi}_2, \dots, \bar{\xi}_m)$  is the Lagrange multiplier which appears in the  $(h, \varphi)$ -Karush-Kuhn-Tucker conditions (7-9). Since  $\bar{\xi}_i > 0$  for  $i \in J$ , we may write (11) as

$$\frac{\bar{\xi}_i}{r} e^{\frac{r}{\bar{\xi}_i} \varphi(g_i(x))} \geq \frac{\bar{\xi}_i}{r} e^{\frac{r}{\bar{\xi}_i} \varphi(g_i(\bar{x}))} \left( 1 + \frac{r}{\bar{\xi}_i p} \bar{\xi}_i h(\nabla^* g_i(\bar{x})) (e^{p\eta(x, \bar{x})} - 1) \right), \quad i \in J. \quad (12)$$

Dividing both sides of inequality (10) by  $e^{r\varphi(f(\bar{x}))}$  and (12) by  $e^{r\varphi(g_i(\bar{x}))}$ , we get

$$\begin{aligned} \frac{1}{r} e^{r(\varphi(f(x)) - \varphi(f(\bar{x})))} &\geq \frac{1}{r} \left( 1 + \frac{r}{p} h(\nabla^* f(\bar{x})) (e^{p\eta(x, \bar{x})} - 1) \right), \\ \frac{\bar{\xi}_i}{r} e^{\frac{r}{\bar{\xi}_i} (\varphi(g_i(x)) - \varphi(g_i(\bar{x})))} &\geq \frac{\bar{\xi}_i}{r} \left( 1 + \frac{r}{\bar{\xi}_i p} \bar{\xi}_i h(\nabla^* g_i(\bar{x})) (e^{p\eta(x, \bar{x})} - 1) \right), \quad i \in J. \end{aligned}$$

After adding both sides of the above inequalities, we obtain

$$\begin{aligned} &\frac{1}{r} \left( e^{r(\varphi(f(x)) - \varphi(f(\bar{x})))} + \sum_{i \in J} \bar{\xi}_i e^{\frac{r}{\bar{\xi}_i} (\varphi(g_i(x)) - \varphi(g_i(\bar{x})))} \right) \geq \frac{1}{r} \left( 1 + \sum_{i \in J} \bar{\xi}_i \right) + \\ &+ \frac{1}{p} \left( \left[ h(\nabla^* f(\bar{x})) + \sum_{i \in J} h(\nabla^* g_i(\bar{x})) \right] (e^{p\eta(x, \bar{x})} - 1) \right) = \\ &= \frac{1}{r} \left( 1 + \sum_{i \in J} \bar{\xi}_i \right) + \frac{1}{p} \left( \left[ h(\nabla^* f(\bar{x})) \oplus \bigoplus_{i=1}^m \bar{\xi}_i \otimes \nabla^* g_i(\bar{x}) \right] (e^{p\eta(x, \bar{x})} - 1) \right). \end{aligned}$$

By (7) and the assumption  $h(0) = 0$  we obtain

$$\frac{1}{r} \left( e^{r(\varphi(f(x)) - \varphi(f(\bar{x})))} + \sum_{i \in J} \bar{\xi}_i e^{\frac{r}{\bar{\xi}_i} (\varphi(g_i(x)) - \varphi(g_i(\bar{x})))} \right) \geq \frac{1}{r} \left( 1 + \sum_{i \in J} \bar{\xi}_i \right).$$

Also using (8) and the assumption  $\varphi(0) = 0$  we get

$$\frac{1}{r} e^{r(\varphi(f(x)) - \varphi(f(\bar{x})))} \geq \frac{1}{r} \left( 1 + \sum_{i \in J} \bar{\xi}_i (1 - e^{r\varphi(g_i(\bar{x}))}) \right), \quad i \in J. \quad (13)$$

Let  $r > 0$  (in the case when  $r < 0$  the proof is analogous, one should change only the directions of inequalities below to the opposite one). Since  $x$  is a feasible point for problem (P),  $g_i(x) \leq 0$  for each  $i = 1, 2, \dots, m$ . This, in turn, implies that  $1 - e^{r\varphi(g_i(\bar{x}))} \geq 0$  for each  $i = 1, 2, \dots, m$ . Taking this facts into account in (13) we get

$$e^{r(\varphi(f(x)) - \varphi(f(\bar{x})))} \geq 1 + \sum_{i \in J} \bar{\xi}_i (1 - e^{r\varphi(g_i(\bar{x}))}) \geq 1.$$

Hence, by assumption that  $\varphi$  is a nondecreasing function, we get

$$\varphi(f(x)) \geq \varphi(f(\bar{x})) \Rightarrow f(x) \geq f(\bar{x}),$$

which means that  $\bar{x}$  is an optimal point in problem (P).

The assumption on functions in Theorem 3.1 could also be given in another form. It is enough to assume that the Lagrange function

$$L(x) = f(x)[+] \left[ \sum_{i=1}^m (\bar{\xi}_i [\cdot] g_i(x)) \right]$$

is  $(h, \varphi) - (p, r)$ -invex with respect to  $\eta$  on  $D$ . And so, the following theorem is true.

**THEOREM 3.2.** *Assume that a point  $\bar{x} \in R^n$  is feasible for problem (P) and let the  $(h, \varphi)$ -Karush-Kuhn-Tucker conditions be satisfied at the point  $(\bar{x}, \bar{\xi})$ . If the function*

$$L(x) = f(x)[+] \left[ \sum_{i=1}^m (\bar{\xi}_i [\cdot] g_i(x)) \right]$$

is a  $(h, \varphi) - (p, r)$ -invex function with respect to  $\eta$  at  $\bar{x}$  on  $D$ , then  $\bar{x}$  is a global minimum for problem (P).

*Proof.* As in Theorem 3.1, Theorem 3.2 will be proved only in case when  $p \neq 0, r \neq 0$ . Let  $x$  be any other feasible point for problem (P). By assumption, the function

$$L(x) = f(x)[+] \left[ \sum_{i=1}^m (\bar{\xi}_i [\cdot] g_i(x)) \right]$$

is  $(h, \varphi) - (p, r)$ -pre invex with respect to  $\eta$  at  $\bar{x}$  on  $D$ . Hence the inequality

$$\frac{1}{r} e^{r\varphi(L(x))} \geq \frac{1}{r} e^{r\varphi(L(\bar{x}))} \left( 1 + \frac{r}{p} h(\nabla^* L(\bar{x})) (e^{p\eta(x, \bar{x})} - 1) \right), \quad i \in J, \quad (14)$$

holds for all  $x \in D$ . First of all we write the  $(h, \varphi)$ -differential of the function  $L$  as

$$\nabla^* L(\bar{x}) = \nabla^* f(\bar{x}) \oplus \bigoplus_{i=1}^m \bar{\xi}_i \otimes \nabla^* g_i(\bar{x}),$$

and now by condition (7) and from (14) we obtain the inequality

$$\frac{1}{r} e^{r\varphi(L(x))} \geq \frac{1}{r} e^{r \left( \sum_{i=1}^m \bar{\xi}_i \varphi(g_i(\bar{x})) - \sum_{i=1}^m \bar{\xi}_i \varphi(g_i(x)) \right)},$$

which can be written as

$$f(x) - f(\bar{x}) \geq \bar{\xi} \cdot \varphi(g(\bar{x})) - \bar{\xi} \cdot \varphi(g(x)).$$

Using (8), (9) and the fact that  $x$  is feasible point for problem (P), we get

$$f(x) \geq f(\bar{x}),$$

which means that the point  $\bar{x}$ , is a global minimum for problem (P).

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