

MULTIOBJECTIVE MATHEMATICAL PROGRAMS WITH EQUILIBRIUM CONSTRAINTS: PSEUDOLINEAR CASE

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Abstract. In this paper, we consider pseudolinear multiobjective mathematical programs with equilibrium constraints. We establish necessary and sufficient strong efficient S-stationary conditions for a feasible point without using any constraint qualification. Although, necessary optimality conditions required constraint qualification, but in pseudolinear case there is no requirement of constraint qualification due to its own characterization. Since duality provide lower bound to the objective function therefore it have good advantage, so we propose Mond-Weir type dual models for a pseudolinear multiobjective mathematical program with equilibrium constraints and deduce usual duality results. Furthermore, some examples are presented to illustrate our results.

Keywords: Pseudolinear multiobjective programming; Mathematical programs with equilibrium constraints; Efficient solution and Duality

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1. Introduction

A mathematical programs with equilibrium constraints (MPEC) is a constrained optimization problem where constraints include equilibrium constraints, such as variational inequalities or complementarity conditions. MPECs has various applications

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in process engineering [1], hydro-economic river basin model [2], traffic and telecommunications networks [3, 4]. MPECs are difficult to solve because the feasible region is not necessarily convex or connected, which cause MFCQ to not hold at every feasible point. This leads to linearly dependent active constraints, the set of multipliers being unbounded, and inconsistent linearizations arbitrarily close to a stationary point [5] also Abadie constraint qualification (ACQ) do not satisfy (see, [6]). Scheel and Scholtes [7] discussed several stationarity conditions like Bouligand stationary (B-stationary), weakly stationary (W-stationary), Clarke stationary (C-stationary), strong stationary (S-stationary) in equilibrium constraints sense. Flegel and Kanzow [16, 17] proposed new constraint qualifications for the MPEC and introduced Karush-Kuhn-Tucker type stationary conditions for the MPEC. Ye [6] established the relationship among Mordukhovich stationary (M-stationary) and several other stationary conditions along with necessary and sufficient optimality conditions for the MPEC.

In real-life, the formation of problems are usually multiobjective. Mordukhovich [18] and Bao *et al.* [19] studied multiobjective optimization problems with equilibrium constraints and established necessary optimality conditions. Recently, Zhang *et al.* [20] extended the existing constraint qualifications from a single objective to the multiobjective case and established various stationarity conditions under the proper Pareto sense. Further, Zhang *et al.* [21] established strong Pareto S-stationarity optimality conditions for multiobjective mathematical programs with equilibrium constraints. Pandey and Mishra [8] proposed strong KKT type sufficient optimality conditions for nonsmooth multiobjective semiinfinite MPEC. Further, Mishra *et al.* [9] established the duality results for MPVC. For a more treatment of the probable applications of MPEC and MPVC, we refer [10, 11, 12, 13, 14, 15] and references therein.

In 1967, Kortanek and Evans [22] studied some properties of a class of functions, which are both pseudoconvex as well as pseudoconcave. This class of functions was

later named as pseudolinear functions by Chew and Choo [23] and established first and second order characterizations for this class of functions. Mishra *et al.* [24] characterised the Lagrange multiplier of solution sets of a nonsmooth pseudolinear optimization problem. For a comprehensive survey on pseudolinear functions and their applications, we refer to the monograph of Mishra and Upadhyay [25].

Therefore, in this paper we consider the pseudolinear multiobjective mathematical programs with equilibrium constraints (PMMPEC) as follows:

$$\begin{aligned} \min \quad & f(z) = (f_1(z), \dots, f_p(z)), \\ \text{subject to} \quad & g_i(z) \leq 0 \quad (i = 1, \dots, q), \quad h_i(z) = 0 \quad (i = 1, \dots, r), \\ & G_i(z) \geq 0, \quad H_i(z) \geq 0, \quad G_i(z)H_i(z) = 0 \quad (i = 1, \dots, m), \end{aligned} \quad (1.1)$$

where $f_i : \mathbb{R}^n \rightarrow \mathbb{R}$, $g_i : \mathbb{R}^n \rightarrow \mathbb{R}$, $h_i : \mathbb{R}^n \rightarrow \mathbb{R}$, $G_i : \mathbb{R}^n \rightarrow \mathbb{R}$ and $H_i : \mathbb{R}^n \rightarrow \mathbb{R}^m$ are continuously differentiable pseudolinear functions on \mathbb{R}^n . There are some applications of multiobjective mathematical programs with equilibrium constraints in healthcare management [26] and determining energy and climate market policy [27]. Duality plays an important role in finding the lower bound of objective functions, thus application point of view duality have great importance. Duality of multiobjective programming problems has been given by several authors (see, [28, 29, 30, 14]).

In Sect. 2, we recall some preliminary results and definitions which will be used throughout the paper. In Sect. 3, we establish necessary and sufficient optimality conditions, formulate Mond-Weir type dual models and establish duality results. We also illustrate duality results under suitable examples.

2. Preliminaries

In this section, we recall some basic definitions and results which will be used throughout the paper. Inequalities between two vectors $y, z \in \mathbb{R}^n$, follow the following conventions:

$$y \preceq z \iff y_i \leq z_i, \quad i = 1, \dots, n,$$

$$y \leq z \iff y \preceq z \quad \text{and} \quad y \neq z,$$

$$y < z \iff y_i < z_i, \quad i = 1, \dots, n.$$

Let S denotes the feasible region of the PMMPEC (1.1) is given as:

$$S = \{z \in \mathbb{R}^n : g(z) \leq 0, h(z) = 0, G(z) \geq 0, H(z) \geq 0, G(z)^T H(z) = 0\}.$$

Let z^* denotes the feasible point of the PMMPEC (1.1). For further discussion we need to define index sets as follows:

$$\alpha(z^*) = \{i : G_i(z^*) = 0, H_i(z^*) > 0\}, \quad \gamma(z^*) = \{i : G_i(z^*) > 0, H_i(z^*) = 0\},$$

$$\beta(z^*) = \{i : G_i(z^*) = 0, H_i(z^*) = 0\}, \quad I_f = \{1, 2, \dots, p\},$$

$$I_g(z^*) = \{i : g_i(z^*) = 0\}, \quad I_g = \{1, 2, \dots, q\}, \quad I_h = \{1, 2, \dots, r\},$$

$$I_h(z^*) = \{i : h_i(z^*) = 0\}, \quad I_f^i = \{1, 2, \dots, p\} \setminus \{i\}.$$

Definition 2.1. [32] A feasible point z^* is said to be an efficient solution of the PMMPEC(1.1), if there is no other feasible point z , such that

$$f(z) \leq f(z^*).$$

Definition 2.2. [25] Let $f : S \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ be a differentiable function on an open convex set S . The function f is said to be

(1) Pseudoconvex at $z^* \in S$, if $\forall z \in S$,

$$\langle \nabla f(z^*), z - z^* \rangle \geq 0 \implies f(z) \geq f(z^*),$$

(2) Pseudoconcave at $z^* \in S$, if $\forall z \in S$,

$$\langle \nabla f(z^*), z - z^* \rangle \leq 0 \implies f(z) \leq f(z^*),$$

The function is said to be pseudoconvex (pseudoconcave) on S if it is pseudoconvex (pseudoconcave) at every $z \in S$. Moreover, the function is said to be pseudolinear on S if it is both pseudoconvex and pseudoconcave on S .

More precisely, a differentiable function $f : S \rightarrow \mathbb{R}$ on an open convex subset $S \subseteq \mathbb{R}^n$, is said to be pseudolinear if $\forall z_1, z_2 \in S$, one has

$$\langle \nabla f(z_1), z_2 - z_1 \rangle \geq 0 \Rightarrow f(z_2) \geq f(z_1),$$

and

$$\langle \nabla f(z_1), z_2 - z_1 \rangle \leq 0 \Rightarrow f(z_2) \leq f(z_1).$$

The following characterization of pseudolinear function given by Chew and Choo [23], which is very useful in the derivation of further results.

Theorem 2.1. *Let $f : S \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ be an open convex set. Then, f is a differentiable pseudolinear function on S if and only if $\forall z_1, z_2 \in S$ there exists a function $p : S \times S \rightarrow \mathbb{R}_+$, where \mathbb{R}_+ denotes positive real number, such that*

$$f(z_2) = f(z_1) + p(z_1, z_2) \langle \nabla f(z_1), z_2 - z_1 \rangle.$$

The function p is called proportional function.

3. Necessary and Sufficient Optimality Conditions

Motivated by strong Pareto S-stationary point of Zhang *et al.* [21], we define strong efficient S-stationary point for pseudolinear multiobjective mathematical programs with equilibrium constraints (1.1) as follows.

Definition 3.1. A feasible point z^* is called strong efficient S-stationary point of the PMMPEC (1.1) if there exist multipliers $(\lambda^f, \lambda^g, \lambda^h, \lambda^G, \lambda^H) \in \mathbb{R}_+^p \times \mathbb{R}^q \times \mathbb{R}^r \times \mathbb{R}^m \times \mathbb{R}^m$ satisfying the following conditions

$$\begin{aligned} & \sum_{i=1}^p \lambda_i^f \nabla f_i(z^*) + \sum_{i=1}^q \lambda_i^g \nabla g_i(z^*) + \sum_{i=1}^r \lambda_i^h \nabla h_i(z^*) - \sum_{i=1}^m \lambda_i^G \nabla G_i(z^*) - \sum_{i=1}^m \lambda_i^H \nabla H_i(z^*) = 0, \\ & \lambda_i^f > 0, \lambda_i^g \geq 0, g(z^*)^T \lambda^g = 0, \lambda_i^G = 0 \ (i \in \gamma(z^*)), \lambda_i^G \geq 0 \ (i \in \alpha(z^*)), \\ & \lambda_i^H = 0 \ (i \in \alpha(z^*)), \lambda_i^H \geq 0 \ (i \in \gamma(z^*)), \lambda_i^G \geq 0, \lambda_i^H \geq 0 \ (i \in \beta(z^*)). \end{aligned}$$

In the following result, we present relationship between strong efficient S-stationary point and efficient solution of pseudolinear multiobjective mathematical programs with equilibrium constraints

Theorem 3.1. *A feasible point z^* is an efficient solution of PMMPEC(1.1) if and only if z^* is an strong efficient S-stationary point of PMMPEC(1.1)*

Proof. Let z^* be a strong efficient S-stationary point, then there exist multipliers $(\lambda^f, \lambda^g, \lambda^h, \lambda^G, \lambda^H) \in \mathbb{R}_+^p \times \mathbb{R}^q \times \mathbb{R}^r \times \mathbb{R}^m \times \mathbb{R}^m$ such that the following conditions hold

$$\sum_{i=1}^p \lambda_i^f \nabla f_i(z^*) + \sum_{i=1}^q \lambda_i^g \nabla g_i(z^*) + \sum_{i=1}^r \lambda_i^h \nabla h_i(z^*) - \sum_{i=1}^m \lambda_i^G \nabla G_i(z^*) - \sum_{i=1}^m \lambda_i^H \nabla H_i(z^*) = 0,$$

$$\lambda_i^f > 0, \lambda_i^g \geq 0, g(z^*)^T \lambda^g = 0, \lambda_i^G = 0 (i \in \gamma(z^*)), \lambda_i^H = 0 (i \in \alpha(z^*)),$$

$$\lambda_i^G \geq 0 (i \in \alpha(z^*)), \lambda_i^H \geq 0 (i \in \gamma(z^*)), \lambda_i^G \geq 0, \lambda_i^H \geq 0 (i \in \beta(z^*)).$$

Assume z^* is not an efficient solution. Then, there exist a feasible point $z \neq z^*$ such that $f_i(z) \leq f_i(z^*)$ for all i except at least one k such that $f_k(z) < f_k(z^*)$. Now, from pseudolinearity we have

$$f_i(z) - f_i(z^*) = p_i^f(z, z^*) \langle \nabla f_i(z^*), z - z^* \rangle \leq 0 \quad \forall i \in \{1, \dots, p\} \setminus \{k\} \tag{3.1}$$

$$f_k(z) - f_k(z^*) = p_k^f(z, z^*) \langle \nabla f_k(z^*), z - z^* \rangle < 0, \tag{3.2}$$

$$g_i(z) - g_i(z^*) = p_i^g(z, z^*) \langle \nabla g_i(z^*), z - z^* \rangle \leq 0, \quad i \in I_g(z^*), \tag{3.3}$$

$$h_i(z) - h_i(z^*) = p_i^h(z, z^*) \langle \nabla h_i(z^*), z - z^* \rangle = 0, \quad i \in I_h(z^*), \tag{3.4}$$

$$-G_i(z) + G_i(z^*) = p_i^G(z, z^*) \langle -\nabla G_i(z^*), z - z^* \rangle \leq 0, \quad i \in \alpha(z^*) \cup \beta(z^*), \tag{3.5}$$

$$-H_i(z) + H_i(z^*) = p_i^H(z, z^*) \langle -\nabla H_i(z^*), z - z^* \rangle \leq 0, \quad i \in \gamma(z^*) \cup \beta(z^*). \tag{3.6}$$

Multiplying (3.1)-(3.6) by $\lambda_i^f > 0 (i \in I_f), \lambda_i^g \geq 0 (i \in I_g(z^*)), \lambda_i^h (i \in I_h(z^*)), \lambda_i^G = 0 (i \in \gamma(z^*)), \lambda_i^G \geq 0 (i \in \alpha(z^*) \cup \beta(z^*)), \lambda_i^H = 0 (i \in \alpha(z^*)), \lambda_i^H \geq 0 (i \in \beta(z^*) \cup$

$\gamma(z^*)$), respectively and using the fact that each $p_i > 0$, we get

$$\left\langle \sum_{i=1}^p \lambda_i^f \nabla f_i(z^*) + \sum_{i=1}^q \lambda_i^g \nabla g_i(z^*) + \sum_{i=1}^r \lambda_i^h \nabla h_i(z^*) - \sum_{i=1}^m \lambda_i^G \nabla G_i(z^*) - \sum_{i=1}^m \lambda_i^H \nabla H_i(z^*), z - z^* \right\rangle < 0,$$

which contradicts the stationarity of z^* as $z - z^* \neq 0$. Hence the result.

Conversely, suppose z^* is an efficient solution of PMMPEC(1.1), then from pseudolinearity of the functions there does not exist any feasible point z different from z^* such that the following system has solution.

$$\begin{aligned} \langle \nabla f_i(z^*), z - z^* \rangle &\leq 0, \quad i = \{1, \dots, p\} \setminus \{k\}, \\ \langle \nabla f_i(z^*), z - z^* \rangle &< 0, \quad i = k, \\ \langle \nabla g_i(z^*), z - z^* \rangle &\leq 0, \quad i \in I_g(z^*), \\ \langle \nabla h_i(z^*), z - z^* \rangle &= 0 \\ \langle -\nabla G_i(z^*), z - z^* \rangle &\leq 0 \quad (i \in \alpha(z^*) \cup \beta(z^*)), \\ \langle -\nabla H_i(z^*), z - z^* \rangle &\leq 0 \quad (i \in \gamma(z^*) \cup \beta(z^*)). \end{aligned} \tag{3.7}$$

That is, the inequality system (3.7) has no solution. Therefore, from Tucker theorem [33] and setting

$$\begin{aligned} \lambda_i^f > 0, \quad i \in I = \{1, \dots, p\}, \quad \lambda_i^g \geq 0, \quad g(z^*)^T \lambda^g = 0, \quad \lambda_i^G = 0 \quad (i \in \gamma(z^*)), \\ \lambda_i^H = 0 \quad (i \in \alpha(z^*)), \quad \lambda_i^G \geq 0 \quad (i \in \alpha(z^*)), \quad \lambda_i^H \geq 0 \quad (i \in \gamma(z^*)), \\ \lambda_i^G \geq 0, \quad \lambda_i^H \geq 0 \quad (i \in \beta(z^*)), \end{aligned}$$

we find $\lambda = (\lambda^f, \lambda^g, \lambda^h, \lambda^H, \lambda^G) \in \mathbb{R}_+^p \times \mathbb{R}^q \times \mathbb{R}^r \times \mathbb{R}^m \times \mathbb{R}^m$, such that

$$\begin{aligned} \sum_{i=1}^p \lambda_i^f \nabla f_i(z^*) + \sum_{i=1}^q \lambda_i^g \nabla g_i(z^*) + \sum_{i=1}^r \lambda_i^h \nabla h_i(z^*) \\ - \sum_{i=1}^m \lambda_i^G \nabla G_i(z^*) - \sum_{i=1}^m \lambda_i^H \nabla H_i(z^*) = 0, \end{aligned} \tag{3.8}$$

Hence, we get the required result. \square

Example 3.1. Consider the problem

$$\min(f_1(z), f_2(z)), \text{ where } f_1(z) = z_1 + z_1^3, f_2(z) = z_2 + z_2^3$$

$$\text{subject to } G(z) = z_1 \geq 0, H(z) = z_2 \geq 0, G(z)H(z) = z_1z_2 = 0.$$

Since,

$$\lambda_1^f \nabla f_1(z) + \lambda_2^f \nabla f_2(z) - \lambda^G \nabla G(z) - \lambda^H \nabla H(z)$$

$$= \lambda_1^f \begin{bmatrix} 1 + 3z_1^2 \\ 0 \end{bmatrix} + \lambda_2^f \begin{bmatrix} 0 \\ 1 + 3z_2^2 \end{bmatrix} - \lambda^G \begin{bmatrix} 1 \\ 0 \end{bmatrix} - \lambda^H \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix},$$

for $\lambda_1^f > 0, \lambda_2^f > 0, \lambda^G = \lambda_1^f(1 + 3z_1^2), \lambda^H = \lambda_2^f(1 + 3z_2^2)$ at point $z = (0, 0)$. Hence, point $z = (0, 0)$ is an efficient solution. Since, except point $z = (0, 0)$ no other point satisfies above conditions (strong efficient S-stationary conditions). That is, there is no efficient solution except $z = (0, 0)$.

Remark 3.1. In Definition 3.1 if we set $\bar{\lambda}_i^f = \frac{\lambda_i^f}{\sum_{i=1}^p \lambda_i^f}, \bar{\lambda}_i^g = \frac{\lambda_i^g}{\sum_{i=1}^q \lambda_i^g}, \bar{\lambda}_i^h = \frac{\lambda_i^h}{\sum_{i=1}^r \lambda_i^h}, \bar{\lambda}_i^G =$

$\frac{\lambda_i^G}{\sum_{i=1}^m \lambda_i^G}$ and $\bar{\lambda}_i^H = \frac{\lambda_i^H}{\sum_{i=1}^m \lambda_i^H}$. Then, Definition 3.1 is reformulated as:

$$\sum_{i=1}^p \bar{\lambda}_i^f \nabla f_i(z^*) + \sum_{i=1}^q \bar{\lambda}_i^g \nabla g_i(z^*) + \sum_{i=1}^r \bar{\lambda}_i^h \nabla h_i(z^*) - \sum_{i=1}^m \bar{\lambda}_i^G \nabla G_i(z^*) - \sum_{i=1}^m \bar{\lambda}_i^H \nabla H_i(z^*) = 0,$$

$$\bar{\lambda}_i^f > 0, \sum_{i=1}^p \bar{\lambda}_i^f = 1, \bar{\lambda}_i^g \geq 0, g(z^*)^T \bar{\lambda}^g = 0, \lambda_i^G = 0 (i \in \gamma(z^*)), \lambda_i^H = 0 (i \in \alpha(z^*)),$$

$$\lambda_i^G \geq 0 (i \in \alpha(z^*)), \lambda_i^H \geq 0 (i \in \gamma(z^*)), \lambda_i^G \geq 0, \lambda_i^H \geq 0 (i \in \beta(z^*)).$$

This is an alternate form of strong efficient S-stationary conditions.

4. Duality

In this section, we propose Mond-Weir type dual model to a pseudolinear MM-PEC(1.1), which is motivated by [31] and establish weak and strong duality results. Mond-Weir type dual for pseudolinear multiobjective mathematical programs with

equilibrium constraints (MWDPMMPEC) is formulated as follows:

$$\max f(u) \quad (4.1)$$

$$\begin{aligned} \text{subject to } S_{MWD} = \{ & (u, \lambda^f, \lambda^g, \lambda^h, \lambda^G, \lambda^H) : \nabla \Phi(u, \lambda^f, \lambda^g, \lambda^h, \lambda^G, \lambda^H) = 0, \\ & \lambda_i^f > 0, \lambda_i^g \geq 0, g(u)^T \lambda^g \geq 0, h(u) = 0, \lambda^G \geq 0 \ (i \in \alpha(u) \cup \beta(u)), \\ & G(u)^T \lambda^G \leq 0, \lambda^H \geq 0 \ (i \in \gamma(u) \cup \beta(u)), \lambda_i^H = 0 \ (i \in \alpha(u)), \\ & \lambda_i^G = 0 \ (i \in \gamma(u)), H(u)^T \lambda^H \leq 0\}, \end{aligned}$$

where $\Phi(u, \lambda^f, \lambda^g, \lambda^h, \lambda^G, \lambda^H)$

$$= \sum_{i=1}^p \lambda_i^f f_i(u) + \sum_{i=1}^q \lambda_i^g g_i(u) + \sum_{i=1}^r \lambda_i^h h_i(u) - \sum_{i=1}^m \lambda_i^G G_i(u) - \sum_{i=1}^m \lambda_i^H H_i(u),$$

Consider the set,

$$S_{MWD}^u = \{u : (u, \lambda^f, \lambda^g, \lambda^h, \lambda^G, \lambda^H) \in S_{MWD}\}.$$

Theorem 4.1. (Weak Duality) *Let z be a feasible point of the PMMPEC (1.1) and $(u, \lambda^f, \lambda^g, \lambda^h, \lambda^G, \lambda^H)$ be a feasible point of the MWDPMMPEC(4.1). If all given functions are pseudolinear at $u \in S \cup S_{MWD}^u$ and any of the following holds:*

- (a) $\lambda_i^f > 0$ and $f_i(\forall i \in I_f)$ are pseudolinear at $u \in S \cup S_{MWD}^u$;
- (b) $\lambda_i^f > 0$ ($\forall i \in I_f$) and $\sum_{i=1}^p \lambda_i^f f_i(\cdot)$ is pseudolinear at $u \in S \cup S_{MWD}^u$.

Then,

$$f(z) \not\leq f(u). \quad (4.2)$$

Proof. Assume that

$$f(z) \leq f(u),$$

Then,

$$f_i(z) \leq f_i(u), \quad \forall i \in I_f, \text{ except at least one } k, \text{ such that}$$

$$f_k(z) < f_k(u).$$

Multiplying by $\lambda_i^f > 0$ and adding, we get

$$(\lambda^f)^T f(z) < (\lambda^f)^T f(u). \quad (4.3)$$

Using the pseudolinearity assumptions, we get

$$\left\langle \sum_{i=1}^p \lambda_i^f \nabla f_i(u), z - u \right\rangle < 0, \tag{4.4}$$

$$\sum_{i=1}^q \lambda_i^g g_i(z) \leq \sum_{i=1}^q \lambda_i^g g_i(u) \implies \left\langle \sum_{i=1}^q \lambda_i^g \nabla g_i(u), z - u \right\rangle \leq 0, \quad i \in I_g(u), \tag{4.5}$$

$$\sum \lambda_i^h h_i(z) = \sum \lambda_i^h h_i(u) \implies \left\langle \sum \lambda_i^h \nabla h_i(u), z - u \right\rangle = 0, \quad i \in I_h, \tag{4.6}$$

$$-\sum \lambda_i^G G_i(z) \leq -\sum \lambda_i^G G_i(u) \implies \left\langle -\sum \lambda_i^G \nabla G_i(u), z - u \right\rangle \leq 0, \quad \forall i, \tag{4.7}$$

$$-\sum \lambda_i^H H_i(z) \leq -\sum \lambda_i^H H_i(u) \implies \left\langle -\sum \lambda_i^H \nabla H_i(u), z - u \right\rangle \leq 0, \quad \forall i, \tag{4.8}$$

by adding (4.4)-(4.8), we get

$$\left\langle \sum_{i=1}^p \lambda_i^f \nabla f_i(u) + \sum_{i=1}^q \lambda_i^g \nabla g_j(u) + \sum_{i=1}^r \lambda_i^h \nabla h_i(u) - \sum_{i=1}^m \lambda_i^G \nabla G_i(u) - \sum_{i=1}^m \lambda_i^H \nabla H_i(u), z - u \right\rangle < 0,$$

which contradicts the feasibility of u . Hence the theorem. □

Theorem 4.2. (Strong Duality) *Let a feasible solution z^* be efficient solution of the PMMPEC (1.1). If assumptions of weak duality Theorem 4.1 holds. Then, there exist $(\bar{\lambda}^f, \bar{\lambda}^g, \bar{\lambda}^h, \bar{\lambda}^G, \bar{\lambda}^H) \in \mathbb{R}_+^p \times \mathbb{R}^q \times \mathbb{R}^r \times \mathbb{R}^m \times \mathbb{R}^m$ such that $(z^*, \bar{\lambda}^f, \bar{\lambda}^g, \bar{\lambda}^h, \bar{\lambda}^G, \bar{\lambda}^H)$ is an efficient solution of the MWDPMMPEC(4.1) and respective objective values are equal.*

Proof. As feasible point z^* is an efficient solution of the pseudolinear MMPEC (1.1). Then, from Theorem 3.1, there exist $(\bar{\lambda}^f, \bar{\lambda}^g, \bar{\lambda}^h, \bar{\lambda}^G, \bar{\lambda}^H) \in \mathbb{R}_+^p \times \mathbb{R}^q \times \mathbb{R}^r \times \mathbb{R}^m \times \mathbb{R}^m$ such that strong efficient S-stationary conditions are satisfied. That is,

$$\sum_{i=1}^p \bar{\lambda}_i^f \nabla f_i(z^*) + \sum_{i=1}^q \bar{\lambda}_i^g \nabla g_i(z^*) + \sum_{i=1}^r \bar{\lambda}_i^h \nabla h_i(z^*) - \sum_{i=1}^m \bar{\lambda}_i^G \nabla G_i(z^*) - \sum_{i=1}^m \bar{\lambda}_i^H \nabla H_i(z^*) = 0.$$

$\bar{\lambda}_i^f > 0, \bar{\lambda}_i^g \geq 0, g(z^*)^T \bar{\lambda}^g = 0, \bar{\lambda}_i^G = 0 (i \in \gamma(z^*)), \bar{\lambda}_i^H = 0 (i \in \alpha(z^*)), \bar{\lambda}_i^G \geq 0, \bar{\lambda}_i^H \geq 0 (i \in \beta(z^*)), \lambda_i^G \geq 0 (i \in \alpha(z^*)), \lambda_i^H \geq 0 (i \in \gamma(z^*)).$ Since z^* is feasible

point then $h_i(z^*) = 0$ ($i \in I_h$), $(\bar{\lambda}^G)^T G(z^*) = 0$, and $(\bar{\lambda}^H)^T H(z^*) = 0$. Therefore, $(z^*, \bar{\lambda}^f, \bar{\lambda}^g, \bar{\lambda}^h, \bar{\lambda}^G, \bar{\lambda}^H)$ is feasible of the MWDPMMPEC(4.1). Then, from feasibility and weak duality Theorem 4.1, we have

$$f(z^*) \geq f(u),$$

for any feasible solution $(u, \lambda^f, \lambda^g, \lambda^h, \lambda^G, \lambda^H) \in \mathbb{R}^n \times \mathbb{R}_+^p \times \mathbb{R}_+^q \times \mathbb{R}^{r+2m}$ of pseudo-linear MWDPMMPEC (4.1). Hence $(z^*, \bar{\lambda}^f, \bar{\lambda}^g, \bar{\lambda}^h, \bar{\lambda}^G, \bar{\lambda}^H)$ is an efficient solution of MWDPMMPEC(4.1) and their respective values are equal. \square

Following example illustrate the Theorem 4.1 and Theorem 4.2.

Example 4.1. Consider the following PMMPEC problem:

$$\begin{aligned} \min f(z) &= (f_1(z), f_2(z)), \text{ where } f_1(z) = z_1 + z_2, f_2(z) = z_2 + z_2^3, \\ \text{subject to } G(z) &= z_1 \geq 0, H(z) = z_2 \geq 0, G(z)^T H(z) = z_1 z_2 = 0, z \in \mathbb{R}^2. \end{aligned}$$

Feasible region for PMMPEC is $S = \{(z_1, z_2) \in \mathbb{R}^2 : z_1 \geq 0, z_2 \geq 0, z_1 z_2 = 0\}$. Now, we formulate MWDPMMPEC dual model according as above discussion.

$$\begin{aligned} \max f(u) &= (u_1 + u_2, u_2 + u_2^3), u \in \mathbb{R}^2, \\ \text{subject to } \lambda_1^f \nabla f_1(u) &+ \lambda_2^f \nabla f_2(u) - \lambda^G \nabla G(u) - \lambda^H \nabla H(u) \\ &= \lambda_1^f \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \lambda_2^f \begin{bmatrix} 0 \\ 1 + 3u_2^2 \end{bmatrix} - \lambda^G \begin{bmatrix} 1 \\ 0 \end{bmatrix} - \lambda^H \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \\ \lambda_1^f &> 0, \lambda_2^f > 0, \lambda^G \geq 0, \lambda^H \geq 0, \lambda^G G(u) = \lambda^G u_1 \leq 0, \\ \lambda^H H(u) &= \lambda^H u_2 \leq 0. \end{aligned}$$

Solving above, we get

$$\lambda^H = \lambda_1^f + \lambda_2^f (1 + 3u_2^2) > 0, \lambda^G = \lambda_1^f > 0 \implies u_1 \leq 0, u_2 \leq 0.$$

It is clear from the feasibility

$$f(z) \not\leq f(u).$$

Hence the weak duality Theorem 4.1 is verified. Now, at point $z^* = (0, 0)$ strong duality Theorem 4.2 can be easily verified.

Conclusions

Kruk and Wolkowicz [34] concluded from Mathis and Mathis [35], that the hospital management problem described by Mathis and Mathis [35] comes under the class of linear constrained pseudolinear optimization problems. The paper of Kruk and Wolkowicz justifies the importance of the class of pseudolinear functions and related programming problems. Since in this paper, we have considered the pseudolinear multiobjective mathematical programs with equilibrium constraints and established necessary and sufficient optimality conditions without any constraint qualifications, proposed Mond-Weir type dual models for pseudolinear multiobjective mathematical programs with equilibrium constraints and established weak and strong duality results with suitable examples. Therefore, application point of view study of pseudolinear multiobjective mathematical programs with equilibrium constraints is useful. Wolfe dual model can be formulated also by taking motivation from [33]. Since wolfe duality results required convexity assumptions, but here all used functions are pseudolinear, which is bigger class. We tried to establish Wolfe duality results, but pseudolinearity arises issues during the proof of weak and strong duality theorems.

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Conflict of interest

The authors declare that they have no conflict of interest.

Data availability statement

I confirm, data will be made available on reasonable request.

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