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Efficiency and Generalized Convex Duality for Multiobjective Programs

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The concept of efficiency (Pareto optimum) is used to formulate duality for multiobjective non-linear programs. The results are obtained for convex functions, ρ -convex (i.e., weakly convex and strongly convex) functions, and pseudoconvex functions. For the convex and ρ -convex functions a Wolfe type of dual is formulated, while for the pseudoconvex and ρ -convex functions, a Mond-Weir type dual is proposed. © 1989 Academic Press, Inc.

1. INTRODUCTION AND PRELIMINARIES

The aim of this paper is to use the concept of efficiency (Pareto optimum) to formulate duality relationships between the multiobjective non-linear program

(VOP) Minimize
$$(f_1(x), f_2(x), ..., f_p(x))$$

Subject to $g(x) \le 0;$ (1)

and the two dual multiobjective programs

Wolfe Vector Dual [11]

(WVD) Maximize
$$(f_1(u) + y'g(u), ..., f_p(u) + y'g(u))$$

Subject to
$$\sum_{i=1}^{p} \tau_i \nabla f_i(u) + \nabla y' g(u) = 0,$$
 (2)

$$y \ge 0, \tag{3}$$

$$\tau_i \ge 0, i = 1, 2, ..., p, \sum_{i=1}^{p} \tau_i = 1;$$
 (4)

0022-247X/89 \$3.00 Copyright © 1989 by Academic Press, Inc. All rights of reproduction in any form reserved. Mond-Weir Vector Dual [6]

(DVOP) Maximize
$$(f_1(u), f_2(u), ..., f_p(u))$$

Subject to (2) through to (4) and

$$y'g(u) \ge 0. \tag{5}$$

The functions $f_i: \mathbb{R}^n \to \mathbb{R}$, i = 1, 2, ..., p and $g: \mathbb{R} \to \mathbb{R}^n$ are assumed to be differentiable.

Optimization in (VOP), (WVD), and (DVOP) means obtaining efficient solutions for the corresponding programs.

DEFINITION 1. A feasible solution x^0 for (VOP) is efficient for (VOP) if and only if there is no other feasible x for (VOP) such that for some $i \in P = \{1, 2, ..., p\}$

$$f_i(x) < f_i(x^0) \tag{6}$$

and

 $f_i(x) \leq f_i(x^0)$ for all $j \in P$. (7)

In the case of maximization, the signs of the inequalities (6) and (7) are reversed (i.e., they become > and \geq , respectively).

For the case where τ_i , i=1, 2, ..., p, are all strictly positive, f_i , i=1, 2, ..., p, and g are convex; Weir [9] has used proper efficiency [3] to establish some duality results between (VOP) and (WVD). Also, Egudo [2] and Weir [9] have used proper efficiency to obtain duality relationships between (VOP) and (DVOP) where the multipliers τ_i , i=1, 2, ..., p, are all strictly positive and a positive linear combination of the objective function components is assumed to be pseudoconvex. Recently Weir [10] obtained a duality result between (VOP) and (DVOP) whereby the objective function components are pseudoconvex but their positive linear combination need not be pseudoconvex.

In [2, 9, 10], the proofs of strong duality results use the following Geoffrion [3] characterization of proper efficiency.

LEMMA 1. If for some fixed $\lambda \in 0 \in \mathbb{R}^p$, x^0 solves the single objective program

 $(P_{\lambda}) \qquad \qquad Minimize \ \lambda' f(x)$ Subject to $g(x) \le 0;$

then x^0 is properly efficient for (VOP).

In this paper proofs for strong duality results will invoke the following.

LEMMA 2 (Theorem 4.1 of Chankong and Haimes [1]). x^0 is an efficient solution for (VOP) if and only if x^0 solves

 $P_{k}(\varepsilon^{0}) \qquad Minimize f_{k}(x)$ Subject to $f_{j}(x) \leq f_{j}(x^{0}) \qquad for \ all \quad j \neq k,$ $g(x) \leq 0;$

for each k = 1, 2, ..., p.

2. WOLFE VECTOR DUALITY

Here we prove weak and strong duality relations between (VOP) and (WVD). First we consider a weak duality result when the functions are convex.

THEOREM 1. Assume that for all feasible x for (VOP) and all feasible (u, τ, y) for (WVD), f_i , i = 1, 2, ..., p, and g are convex functions. If also either

(a)
$$\tau_i > 0$$
 for all $i = 1, 2, ..., p$ or

(b) $\sum_{i=1}^{p} \tau_i f_i(\cdot) + \sum_{i=1}^{m} y_i g_i(\cdot)$ is strictly convex at u,

then the following cannot hold:

$$f_j(x) \le f_j(u) + y'g(u)$$
 for all $j \in P = \{1, 2, ..., p\}$ (8)

and

$$f_i(x) < f_i(u) + y'g(u)$$
 for some $i \in P$. (9)

Proof. Suppose contrary to the result that (8) and (9) hold. Then since x is feasible for (VOP) and $y \ge 0$, (8) and (9) imply

$$f_j(x) + y'g(x) \le f_j(u) + y'g(u) \quad \text{for all} \quad j \in P \tag{10}$$

and

$$f_i(x) + y'g(x) < f_i(u) + y'g(u) \quad \text{for some} \quad i \in P,$$
(11)

respectively. Now hypothesis (a) and $\sum_{i=1}^{p} \tau_i = 1$ imply

$$\sum_{i=1}^{p} \tau_i f_i(x) + y^i g(x) < \sum_{i=1}^{p} \tau_i f_i(u) + y^i g(u)$$
(12)

and since f's and g are convex and $\tau_i > 0$, $i = 1, 2, ..., p, y \ge 0$, it now follows from (12) that

$$(x-u)'\left(\sum_{i=1}^{p}\tau_i\nabla f_i(u)+\nabla y'g(u)\right)<0,$$
(13)

which contradicts (2).

Also, since $\tau_i \ge 0$, i = 1, 2, ..., p, and $\sum_{i=1}^{p} \tau_i = 1$, (10) and (11) imply

$$\sum_{i=1}^{p} \tau_i f_i(x) + y^i g(x) \leq \sum_{i=1}^{p} \tau_i f_i(u) + y^i g(u).$$
(14)

Now (14) and hypothesis (b) imply (13), again contradicting (2).

Next we state and prove a weak duality result between (VOP) and (WVD) under ρ -convexity. But first we define ρ -convex functions [7, 8].

DEFINITION 2. A function $f: \mathbb{R}^n \to \mathbb{R}$ is said to be ρ -convex [7, 8] if there exists a real number ρ such that for each $x, u \in \mathbb{R}^n$ and $0 \leq \lambda \leq 1$,

$$f(\lambda x + (1 - \lambda)u) \leq \lambda f(x) + (1 - \lambda) f(u) - \rho \lambda (1 - \lambda) ||x - u||^2.$$

For a differentiable function $f: \mathbb{R}^n \to \mathbb{R}$, f is ρ -convex if and only if for all x, $u \in \mathbb{R}^n$

$$f(x) - f(u) \ge (x - u)^t \nabla f(u) + \rho ||x - u||^2$$
.

If ρ is positive then f is said to be strongly convex [7, 8]; and if ρ is negative then f is said to be weakly convex [8].

THEOREM 2 (Weak Duality). Assume that for all feasible x for (VOP) and all feasible (u, τ, y) for (WVD), f_i , i = 1, 2, ..., p, are ρ_i -convex and g_j , j = 1, 2, ..., m, are σ_j -convex. If also either

- (a) $\tau_i > 0$ for all i = 1, 2, ..., p and $\sum_{i=1}^{p} \tau_i \rho_i + \sum_{i=1}^{m} y_i \sigma_i \ge 0$ or
- (b) $\sum_{i=1}^{p} \tau_i \rho_i + \sum_{i=1}^{m} y_i \sigma_i > 0$,

then the following cannot hold:

$$f_j(x) \le f_j(u) + y^t g(u)$$
 for all $j \in P = \{1, 2, ..., p\}$ (15)

and

$$f_i(x) < f_i(u) + y^t g(u)$$
 for some $i \in P$. (16)

Remark 1. Hypothesis (a) can be interpreted as follows: when all the objective function multipliers are strictly positive then the linear com-

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bination of the objective function components plus the non-negative linear combination of the constraint functions should be either convex or strongly convex. Hypothesis (b) can be interpreted as follows: the non-negative linear combination of the objective function components and the constraint functions should be strongly convex. These conditions are weaker than those in Theorem 1 because they allow some of the objective function components and constraint functions to be weakly convex [8] provided that their non-negative linear combination is either convex or strongly convex.

Proof. Suppose contrary to the result that (15) and (16) hold. Then since x is feasible for (VOP) and $y \ge 0$, (15) and (16) imply

$$f_j(x) + y'g(x) \le f_j(u) + y'g(u) \quad \text{for all} \quad j \in P \tag{17}$$

and

$$f_i(x) + y'(x) < f_i(u) + y'g(u) \quad \text{for some} \quad i \in P.$$
(18)

Now if hypothesis (a) holds, then from $\tau_i > 0$ for all $i \in P$, (17), and (18) we obtain

$$\sum_{i=1}^{p} \tau_{i} f_{i}(x) + y^{t} g(x) \sum_{i=1}^{p} \tau_{i} < \sum_{i=1}^{p} \tau_{i} f_{i}(u) + y^{t} g(u) \sum_{i=1}^{p} \tau_{i}$$

and since $\sum_{i=1}^{p} \tau_i = 1$, this inequality reduces to

$$\sum_{i=1}^{p} \tau_i(f_i(x) - f_i(u)) + y'g(x) - y'g(u) < 0.$$
(19)

Now from (19), ρ_i -convexity of f_i 's, and σ_i -convexity of g_j 's we obtain

$$(x-u)^{\prime}\left(\sum_{i=1}^{p}\tau_{i}\nabla f_{i}(u)+\nabla y^{\prime}g(u)\right)$$
$$+\left(\sum_{i=1}^{p}\tau_{i}\rho_{i}+\sum_{j=1}^{m}y_{j}\sigma_{j}\right)\|x-u\|^{2}<0$$

and since by hypothesis (a), $\sum_{i=1}^{p} \tau_i \rho_i + \sum_{j=1}^{m} y_j \sigma_j \ge 0$, this implies

$$(x-u)^{t}\left(\sum_{i=1}^{p}\tau_{i}\nabla f_{i}(u)+y^{t}\nabla g(u)\right)<0,$$
(20)

which contradicts (2).

Also from (17), (18), and $\tau_i \ge 0$, i = 1, 2, ..., p, $\sum_{i=1}^{p} \tau_i = 1$, we obtain

$$\sum_{i=1}^{p} \tau_i(f_i(x) - f_i(u)) + y'g(x) - y'g(u) \le 0;$$
(21)

and since f_i 's are ρ_i -convex and g_i 's are σ_i -convex, (21) implies

$$(x-u)^{t}\left(\sum_{i=1}^{p}\tau_{i}\nabla f_{i}(u)+\nabla y^{t}g(u)\right)$$
$$+\left(\sum_{i=1}^{p}\tau_{i}\rho_{i}+\sum_{j=1}^{m}y_{j}\sigma_{j}\right)\|x-u\|^{2} \leq 0.$$
(22)

Now by hypothesis (b), $\sum_{i=1}^{p} \tau_i \rho_i + \sum_{j=1}^{m} y_j \sigma_j > 0$; hence (22) implies (20), again contradicting (2).

COROLLARY 1. Assume that weak duality (Theorem 1 or 2) holds between (VOP) and (WVD). If (u^0, τ^0, y^0) is feasible for (WVD) with $y^{0t}g(u^0) = 0$ and u^0 is feasible for (VOP), then u^0 is efficient for (VOP) and (u^0, τ^0, y^0) is efficient for (WVD).

Proof. Suppose that u^0 is not efficient for (VOP); then there exists a feasible x for (VOP) such that for some $i \in P = \{1, 2, ..., p\}$,

$$f_i(x) < f_i(u^0) \tag{23}$$

and

$$f_j(x) \le f_j(u^0)$$
 for all $j \in P$. (24)

By hypothesis $y^{0t}g(u^0) = 0$, so (23) and (24) can be written as

$$f_u(x) < f_i(u^0) + y^{0i}g(u^0) \quad \text{for some} \quad i \in P$$

and $f_j(x) \le f_j(u^0) + y^{0i}g(u^0) \quad \text{for all} \quad j \in P$,

respectively; and since (u^0, τ^0, y^0) is feasible in (WVD) and x is feasible for (VOP), these inequalities contradict weak duality (Theorem 1 or 2).

Also suppose that (u^0, τ^0, y^0) is not efficient of (WVD). Then there exists a feasible (u, τ, y) for (WVD) such that for some $i \in P = \{1, 2, ..., p\}$

$$f_i(u) + y'g(u) > f_i(u^0) + y^{0'}g(u^0)$$
(25)

and

$$f_j(\boldsymbol{u}) + y^{\prime}g(\boldsymbol{u}) \ge f_j(\boldsymbol{u}^0) + y^{0\prime}g(\boldsymbol{u}^0) \quad \text{for all} \quad j \in \boldsymbol{P};$$
(26)

and since $y^{0t}g(u^0) = 0$, (25) and (26) reduce to

$$f_i(u) + y'g(u) > f_i(u^0)$$
 for some $i \in P$

and

$$f_j(u) + y'g(u) \ge f_j(u^0)$$
 for all $j \in P$,

respectively. Since u^0 is feasible for (VOP), these inequalities contradict weak duality (Theorem 1 or 2). Therefore u^0 and (u^0, τ^0, y^0) are efficient for their respective programs.

THEOREM 3 (Strong Duality). Let x^0 be an efficient solution for (VOP) and assume that x^0 satisfies a constraint qualification [4, 5] for $P_k(\varepsilon^0)$ for at least one k = 1, 2, ..., p; then there exist $\tau^0 \in \mathbb{R}^p$ and $y^0 \in \mathbb{R}^m$ such (x^0, τ^0, y^0) is feasible for (WVD) and $y^{0t}g(x^0) = 0$. If also weak duality (Theorem 1 or 2) holds between (VOP) and (WVD) then (x^0, τ^0, y^0) is efficient for (WVD).

Proof. Since x^0 is efficient for (VOP), from Lemma 2, x^0 solves $P_k(\varepsilon^0)$ for all k = 1, 2, ..., p. By hypothesis there exists a $k \in P = \{1, 2, ..., p\}$ for which x^0 satisfies a constraint qualification of $P_k(\varepsilon^0)$. Now from Kuhn-Tucker necessary conditions [4, 5] there exist $\tau_i \ge 0$ for all $i \ne k$ and $y \ge 0 \in \mathbb{R}^m$ such that

$$f_k(x^0) + \sum_{i \neq k} \tau_i \nabla f_i(x^0) + \sum_{j=1}^m y_j \nabla g_j(x^0) = 0,$$
(27)

$$y'g(x^0) = 0.$$
 (28)

Now dividing all terms in (27) and (28) by $1 + \sum_{i \neq k} \tau_i$ and setting

$$\tau_k^0 = \frac{1}{1 + \sum_{i \neq k} \tau_i} > 0, \qquad \tau_j^0 = \frac{\tau_j}{j + \sum_{1 \neq k} \tau_i} \ge 0,$$

and

$$y^0 = \frac{y}{1 + \sum_{i \neq k} \tau_i} \ge 0,$$

we conclude that (x^0, τ^0, y^0) is feasible for (WVD). Efficiency of (x^0, τ^0, y^0) for (WVD) now follows from Corollary 1.

3. MOND-WEIR VECTOR DUALITY

In this section we present weak and strong duality relations between programs (VOP) and (DVOP). The weak duality results are given under two conditions: one when objective function components are pseudoconvex and the constraint functions are quasiconvex and the other when the functions are ρ -convex.

THEOREM 4 (Weak Duality). Assume that for all feasible X for (VOP) and all feasible (u, τ, y) for (DVOP), $y'g(\cdot)$ is quasiconvex at u. If also any of the following holds.

(a) $\tau_i > 0, \forall i \in P = \{1, 2, ..., p\}, and f_i, i = 1, 2, ..., p, are pseudoconvex at u;$

- (b) $\tau_i > 0$, for all $i \in P$ and $\sum_{i=1}^{p} \tau_i f_i(\cdot)$ is pseudoconvex at u;
- (c) $\sum_{i=1}^{p} \tau_i f_u(\cdot)$ is strictly pseudoconvex at u,

then the following cannot hold:

$$f_j(x) \leq f_j(u)$$
 for all $j \in P = \{1, 2, ..., p\}$ (29)

and

$$f_j(x) < f_i(u)$$
 for some $i \in P$. (30)

Proof. For each feasible x for (VOP) and each feasible (u, τ, y) for (DVOP) we have $y'g(x) - y'g(u) \leq 0$; and since $y'g(\cdot)$ is quasiconvex at u this implies

$$(x-y)' \nabla y' g(u) \leq 0. \tag{31}$$

Applying (31) to $\sum_{i=1}^{p} \tau_i \nabla f_i(u) + \nabla y' g(u) = 0$ yields

$$(x-u)^t \sum_{i=1}^p \tau_i \nabla f_i(u) \ge 0.$$
(32)

Now suppose contrary to the result of the theorem that (29) and (30) hold. If $\tau_i > 0$ for all i = 1, 2, ..., p then (29) and (30) imply

$$\tau_j f_j(x) \le \tau_j f_j(u)$$
 for all $j \in P = \{1, 2, ..., p\}$ (33)

and

$$\tau_i f_i(x) < \tau_i f_i(u)$$
 for some $i \in P$, (34)

respectively. Equations (33) and (34) also imply

$$\sum_{i=1}^{p} \tau_i f_i(x) < \sum_{i=1}^{p} \tau_i f_i(u).$$
(35)

By hypothesis (a), i.e., f_i 's are pseudoconvex, (33) and (34) imply

$$(x-u)'\left(\sum_{i=1}^{p}\tau_i\nabla f_i(u)\right) < 0, \tag{36}$$

contradicting (32).

By hypothesis (b), i.e., $\sum_{i=1}^{p} \tau_i f_i(\cdot)$ is pseudoconvex at u, (35) implies (36), again contradicting (32).

Now from $\tau_i \ge 0$, i = 1, 2, ..., p, (29), and (30) we obtain

$$\sum_{i=1}^{p} \tau_{i} f_{i}(x) \leq \sum_{i=1}^{p} \tau_{i} f_{i}(u)$$
(37)

and by hypothesis (c), i.e., $\sum_{i=1}^{p} \tau_i f_i(\cdot)$ is strictly pseudoconvex, (37) implies (36), again contradicting (32).

THEOREM 5 (Weak Duality). Assume that for all feasible x for (VOP) and all feasible (u, τ, y) for (DVOP), f_i , i = 1, 2, ..., p, are ρ_i -convex and g_j , j = 1, 2, ..., m, are σ_j -convex. If either

(a) $\tau_i > 0$ for all $i \in P = \{1, 2, ..., p\}$ and $\sum_{i=1}^{p} \tau_i \rho_i + \sum_{j=1}^{m} y_j \sigma_j \ge 0$, or (b) $\sum_{i=1}^{p} \tau_i \rho_i + \sum_{j=1}^{m} y_j \sigma_j > 0$,

then the following cannot hold:

$$f_i(x) < f_i(u)$$
 for some $i \in P = \{1, 2, ..., p\}$ (38)

and

$$f_i(x) \leq f_i(u) \quad \text{for all} \quad j \in P.$$
 (39)

Proof. Suppose contrary to the result that (38) and (39) hold; then for $\tau_i > 0$ for each i = 1, 2, ..., p, (38) and (39) imply

$$\sum_{i=1}^{p} \tau_i f_i(x) < \sum_{i=1}^{p} \tau_i f_i(u);$$
(40)

and for $\tau_i \ge 0$, i = 1, 2, ..., p, (38) and (39) imply

$$\sum_{i=1}^{p} \tau_{i} f_{i}(x) \leq \sum_{i=1}^{p} \tau_{i} f_{i}(u).$$
(41)

Now since f's are ρ_i -convex (40) and (41) imply

$$(x-u)^{t}\left(\sum_{i=1}^{p}\tau_{i}\nabla f_{i}(u)\right)+\left(\sum_{i=1}^{p}\tau_{i}\rho_{i}\right)\|x-u\|^{2}<0,$$
(42)

$$(x-u)^{\prime} \left(\sum_{i=1}^{p} \tau_{i} \nabla f_{i}(u) \right) + \left(\sum_{i=1}^{p} \tau_{i} \rho_{i} \right) \|x-u\|^{2} \leq 0,$$
(43)

respectively. Also, since (u, τ, y) is feasible for (DVOP) and x is feasible for (VOP) we have

$$y'g(x) - y'g(u) \leq 0; \tag{44}$$

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and because g_i 's are σ_i -convex (44) implies

$$(x-u)' \nabla y' g(u) + ||x-u||^2 \sum_{j=1}^m y_j \sigma_j \leq 0.$$
 (45)

Now adding (42) and (45) and then applying hypothesis (a), i.e.,

$$\sum_{i=1}^{p} \tau_i \rho_i + \sum_{j=1}^{m} y_j \sigma_j \ge 0,$$

yield

$$(x-u)^{t}\left(\sum_{i=1}^{p}\tau_{i}\nabla f_{i}(u)+\sum_{j=1}^{m}y_{j}\nabla g_{j}(u)\right)<0,$$
(46)

which contradicts (2). Also, adding (43) and (45) and then applying hypothesis (b) yield (46), again contradicting (2). \blacksquare

COROLLARY 2. Assume weak duality (Theorem 4 or 5) holds between (VOP) and (DVOP). If (u^0, τ^0, y^0) is feasible for (DVOP) such that u^0 is feasible for (VOP), then u^0 is efficient for (VOP) and (u^0, τ^0, y^0) is efficient for (DVOP).

Proof. First we show that u^0 is efficient for (VOP). Suppose that u^0 is not efficient for (VOP); then there exists a feasible x for (VOP) such that (29) and (30) or (38) and (39) hold. But (u^0, τ^0, y^0) is feasible for (DVOP); hence the result of weak duality (Theorem 4 or 5) is contradicted. Therefore u^0 must be efficient for (VOP). Similarly assuming that (u^0, τ^0, y^0) is not efficient for (DVOP) leads to a contradiction, and hence (u^0, τ^0, y^0) is efficient for (DVOP).

THEOREM 6 (Strong Duality). Let x^0 be efficient for (VOP) and assume that x^0 satisfies a constraint qualification [4, 5] for $P_k(\varepsilon^0)$ for at least one k = 1, 2, ..., p. Then there exist $\tau^0 \in \mathbb{R}^p$ and $y^0 \in \mathbb{R}^m$ such that (x^0, τ^0, y^0) is feasible for (DVOP). If also weak duality (Theorem 4 or 5) holds between (VOP) and (DVOP) then (x^0, τ^0, y^0) is efficient for (DVOP).

Proof. Since x^0 is an efficient solution of (VOP), then from Lemma 2, x^0 solves $P_k(\varepsilon^0)$ for each k = 1, 2, ..., p. By hypothesis there exists at least one k = 1, 2, ..., p such that x^0 satisfies a constraint qualification [4, 5] for $P_k(\varepsilon^0)$. From Kuhn-Tucker necessary conditions [4, 5] we obtain $\tau_i \ge 0$ for all $i \ne k$, and $y \ge 0 \in \mathbb{R}^m$ such that

$$\nabla f_k(x^0) + \sum_{i \neq k} \tau_i \nabla f_i(x^0) + \nabla y' g(x^0) = 0$$
(47)

$$y'g(x^0) = 0.$$
 (48)

Now dividing (47) and (48) by $1 + \sum_{i \neq k} \tau_i$ and defining

$$\tau_k^0 = \frac{1}{1 + \sum_{i \neq k} \tau_i} > 0, \quad \tau_j^0 = \frac{\tau_j}{1 + \sum_{i \neq k} \tau_i} \ge 0 \quad \text{for all} \quad j \neq k$$

and

$$y^0 = \frac{y}{1 + \sum_{i \neq k} \tau_i} \ge 0,$$

we conclude that (x^0, τ^0, y^0) is feasible for (DVOP). The efficiency of (x^0, τ^0, y^0) for (DVOP) now follows from Corollary 2.

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