# 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, $\rho$-convex (i.e., weakly convex and strongly convex) functions, and pseudoconvex functions. For the convex and $\rho$-convex functions a Wolfe type of dual is formulated, while for the pseudoconvex and $\rho$-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)

$$
\begin{align*}
& \text { Minimize }\left(f_{1}(x), f_{2}(x), \ldots, f_{p}(x)\right) \\
& \text { Subject to } g(x) \leqq 0 \tag{1}
\end{align*}
$$

and the two dual multiobjective programs
Wolfe Vector Dual [11]
(WVD) $\quad \operatorname{Maximize}\left(f_{1}(u)+y^{\prime} g(u), \ldots, f_{p}(u)+y^{\prime} g(u)\right)$

$$
\begin{align*}
& \text { Subject to } \sum_{i=1}^{p} \tau_{i} \nabla f_{i}(u)+\nabla y^{\prime} g(u)=0,  \tag{2}\\
& y \geqq 0,  \tag{3}\\
& \tau_{i} \geqq 0, i=1,2, \ldots, p, \sum_{i=1}^{p} \tau_{i}=1 ; \tag{4}
\end{align*}
$$

Mond-Weir Vector Dual [6]
(DVOP) Maximize $\left(f_{1}(u), f_{2}(u), \ldots, f_{p}(u)\right)$
Subject to (2) through to (4) and

$$
\begin{equation*}
y^{\prime} g(u) \geqq 0 \tag{5}
\end{equation*}
$$

The functions $f_{i}: \mathbb{R}^{n} \rightarrow \mathbb{R}, i=1,2, \ldots, p$ and $g: \mathbb{R} \rightarrow \mathbb{R}^{n}$ 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 $i \in P=\{1,2, \ldots, p\}$

$$
\begin{equation*}
f_{i}(x)<f_{i}\left(x^{0}\right) \tag{6}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{j}(x) \leqq f_{j}\left(x^{0}\right) \quad \text { for all } \quad j \in P \tag{7}
\end{equation*}
$$

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

For the case where $\tau_{i}, i=1,2, \ldots, p$, are all strictly positive, $f_{i}, i=1$, $2, \ldots, p$, 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 $\tau_{i}, i=1$, $2, \ldots, p$, 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

Minimize $\lambda^{t} f(x)$
Subject to $g(x) \leqq 0$;
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
$P_{k}\left(\varepsilon^{0}\right) \quad$ Minimize $f_{k}(x)$

$$
\begin{gathered}
\text { Subject to } f_{j}(x) \leqq f_{j}\left(x^{0}\right) \quad \text { for all } j \neq k, \\
g(x) \leqq 0
\end{gathered}
$$

for each $k=1,2, \ldots, p$.

## 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 $(u, \tau, y)$ for $(W V D), f_{i}, i=1,2, \ldots, p$, and $g$ are convex functions. If also either
(a) $\tau_{i}>0$ for all $i=1,2, \ldots, p$ or
(b) $\sum_{i=1}^{p} \tau_{i} f_{i}(\cdot)+\sum_{j=1}^{m} y_{j} g_{j}(\cdot)$ is strictly convex at $u$,
then the following cannot hold:

$$
\begin{equation*}
f_{j}(x) \leqq f_{j}(u)+y^{\prime} g(u) \quad \text { for all } j \in P=\{1,2, \ldots, p\} \tag{8}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{i}(x)<f_{i}(u)+y^{t} g(u) \quad \text { for some } \quad i \in P . \tag{9}
\end{equation*}
$$

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

$$
\begin{equation*}
f_{j}(x)+y^{t} g(x) \leqq f_{j}(u)+y^{t} g(u) \quad \text { for all } \quad j \in P \tag{10}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{i}(x)+y^{\prime} g(x)<f_{i}(u)+y^{\prime} g(u) \quad \text { for some } \quad i \in P \tag{11}
\end{equation*}
$$

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

$$
\begin{equation*}
\sum_{i=1}^{p} \tau_{i} f_{i}(x)+y^{t} g(x)<\sum_{i=1}^{p} \tau_{i} f_{i}(u)+y^{t} g(u) \tag{12}
\end{equation*}
$$

and since $f_{i}$ 's and $g$ are convex and $\tau_{i}>0, i=1,2, \ldots, p, y \geqq 0$, it now follows from (12) that

$$
\begin{equation*}
(x-u)^{t}\left(\sum_{i=1}^{p} \tau_{i} \nabla f_{i}(u)+\nabla y^{t} g(u)\right)<0 \tag{13}
\end{equation*}
$$

which contradicts (2).
Also, since $\tau_{i} \geqq 0, i=1,2, \ldots, p$, and $\sum_{i=1}^{p} \tau_{i}=1,(10)$ and (11) imply

$$
\begin{equation*}
\sum_{i=1}^{p} \tau_{i} f_{i}(x)+y^{t} g(x) \leqq \sum_{i=1}^{p} \tau_{i} f_{i}(u)+y^{t} g(u) \tag{14}
\end{equation*}
$$

Now (14) and hypothesis (b) imply (13), again contradicting (2).
Next we state and prove a weak duality result between (VOP) and (WVD) under $\rho$-convexity. But first we define $\rho$-convex functions [7, 8].

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

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

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

$$
f(x)-f(u) \geqq(x-u)^{t} \nabla f(u)+\rho\|x-u\|^{2}
$$

If $\rho$ is positive then $f$ is said to be strongly convex $[7,8]$; and if $\rho$ 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 $(u, \tau, y)$ for $(W V D), f_{i}, i=1,2, \ldots, p$, are $\rho_{i}$-convex and $g_{i}$, $j=1,2, \ldots, m$, are $\sigma_{j}$-convex. If also either
(a) $\tau_{i}>0$ for all $i=1,2, \ldots, p$ and $\sum_{i=1}^{p} \tau_{i} \rho_{i}+\sum_{j=1}^{m} y_{j} \sigma_{j} \geqq 0$ or
(b) $\sum_{i=1}^{p} \tau_{i} \rho_{i}+\sum_{j=1}^{m} y_{j} \sigma_{j}>0$,
then the following cannot hold:

$$
\begin{equation*}
f_{j}(x) \leqq f_{j}(u)+y^{t} g(u) \quad \text { for all } \quad j \in P=\{1,2, \ldots, p\} \tag{15}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{i}(x)<f_{i}(u)+y^{t} g(u) \quad \text { for some } \quad i \in P \tag{16}
\end{equation*}
$$

Remark 1. Hypothesis (a) can be interpreted as follows: when all the objective function multipliers are strictly positive then the linear com-
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 $y \geqq 0$, (15) and (16) imply

$$
\begin{equation*}
f_{j}(x)+y^{\prime} g(x) \leqq f_{j}(u)+y^{\prime} g(u) \quad \text { for all } j \in P \tag{17}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{i}(x)+y^{t}(x)<f_{i}(u)+y^{t} g(u) \quad \text { for some } \quad i \in P . \tag{18}
\end{equation*}
$$

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

$$
\begin{equation*}
\sum_{i=1}^{p} \tau_{i}\left(f_{i}(x)-f_{i}(u)\right)+y^{t} g(x)-y^{\prime} g(u)<0 \tag{19}
\end{equation*}
$$

Now from (19), $\rho_{i}$-convexity of $f_{i}$ 's, and $\sigma_{j}$-convexity of $g_{j}$ 's we obtain

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

and since by hypothesis (a), $\sum_{i=1}^{p} \tau_{i} \rho_{i}+\sum_{j=1}^{m} y_{j} \sigma_{j} \geqq 0$, this implies

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

which contradicts (2).
Also from (17), (18), and $\tau_{i} \geqq 0, i=1,2, \ldots, p, \sum_{i=1}^{p} \tau_{i}=1$, we obtain

$$
\begin{equation*}
\sum_{i=1}^{p} \tau_{i}\left(f_{i}(x)-f_{i}(u)\right)+y^{t} g(x)-y^{t} g(u) \leqq 0 \tag{21}
\end{equation*}
$$

and since $f_{i}^{\prime}$ 's are $\rho_{i}$-convex and $g_{j}$ 's are $\sigma_{j}$-convex, (21) implies

$$
\begin{align*}
& (x-u)^{t}\left(\sum_{i=1}^{n} \tau_{i} \nabla f_{i}(u)+\nabla y^{t} g(u)\right) \\
& \quad+\left(\sum_{i=1}^{p} \tau_{i} \rho_{i}+\sum_{j=1}^{m} y_{j} \sigma_{j}\right)\|x-u\|^{2} \leqq 0 . \tag{22}
\end{align*}
$$

Now by hypothesis (b), $\sum_{i=1}^{p} \tau_{i} \beta_{i}+\sum_{j=1}^{m} y_{j} \sigma_{j}>0$; hence (22) implies (20), again contradicting (2).

Corollary 1. Assume that weak duality (Theorem 1 or 2) holds between (VOP) and (WVD). If $\left(u^{0}, \tau^{0}, y^{0}\right)$ is feasible for (WVD) with $y^{0 t} g\left(u^{0}\right)=0$ and $u^{0}$ is feasible for (VOP), then $u^{0}$ is efficient for (VOP) and ( $u^{0}, \tau^{0}, y^{0}$ ) 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 $i \in P=\{1,2, \ldots, p\}$,

$$
\begin{equation*}
f_{i}(x)<f_{i}\left(u^{0}\right) \tag{23}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{j}(x) \leqq f_{j}\left(u^{0}\right) \quad \text { for all } \quad j \in P \tag{24}
\end{equation*}
$$

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

$$
\begin{aligned}
& f_{u}(x)<f_{i}\left(u^{0}\right)+y^{0} g\left(u^{0}\right) \quad \text { for some } \quad i \in P \\
& \text { and } f_{j}(x) \leqq f_{j}\left(u^{0}\right)+y^{0} g\left(u^{0}\right) \quad \text { for all } \quad j \in P
\end{aligned}
$$

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

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

$$
\begin{equation*}
f_{i}(u)+y^{t} g(u)>f_{i}\left(u^{0}\right)+y^{0} g\left(u^{0}\right) \tag{25}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{j}(u)+y^{t} g(u) \geqq f_{j}\left(u^{0}\right)+y^{0 t} g\left(u^{0}\right) \quad \text { for all } \quad j \in P \tag{26}
\end{equation*}
$$

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

$$
f_{i}(u)+y^{\prime} g(u)>f_{i}\left(u^{0}\right) \quad \text { for some } \quad i \in P
$$

and

$$
f_{j}(u)+y^{t} g(u) \geqq f_{j}\left(u^{0}\right) \quad \text { for all } \quad j \in P
$$

respectively. Since $u^{0}$ is feasible for (VOP), these inequalities contradict weak duality (Theorem 1 or 2 ). Therefore $u^{0}$ and ( $u^{0}, \tau^{0}, y^{0}$ ) 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 $P_{k}\left(\varepsilon^{0}\right)$ for at least one $k=1,2, \ldots, p$; then there exist $\tau^{0} \in \mathbb{R}^{p}$ and $y^{0} \in \mathbb{R}^{m} \operatorname{such}\left(x^{0}, \tau^{0}, y^{0}\right)$ is feasible for ( $W V D$ ) and $y^{0 t} g\left(x^{0}\right)=0$. If also weak duality (Theorem 1 or 2) holds between (VOP) and (WVD) then $\left(x^{0}, \tau^{0}, y^{0}\right)$ is efficient for (WVD).

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

$$
\begin{align*}
f_{k}\left(x^{0}\right)+\sum_{i \neq k} \tau_{i} \nabla f_{i}\left(x^{0}\right)+\sum_{j=1}^{m} y_{j} \nabla g_{j}\left(x^{0}\right) & =0  \tag{27}\\
y^{t} g\left(x^{0}\right) & =0 \tag{28}
\end{align*}
$$

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

$$
\tau_{k}^{0}=\frac{1}{1+\sum_{i \neq k} \tau_{i}}>0, \quad \tau_{j}^{0}=\frac{\tau_{j}}{j+\sum_{1 \neq k} \tau_{i}} \geqq 0
$$

and

$$
y^{0}=\frac{y}{1+\sum_{i \neq k} \tau_{i}} \geqq 0
$$

we conclude that ( $x^{0}, \tau^{0}, y^{0}$ ) is feasible for (WVD). Efficiency of ( $x^{0}, \tau^{0}, y^{0}$ ) 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 $\rho$-convex.

Theorem 4 (Weak Duality). Assume that for all feasible $X$ for (VOP) and all feasible $(u, \tau, y)$ for (DVOP), $y^{\prime} g(\cdot)$ is quasiconvex at $u$. If also any of the following holds.
(a) $\tau_{i}>0, \forall i \in P=\{1,2, \ldots, p\}$, and $f_{i}, i=1,2, \ldots, p$, 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:

$$
\begin{equation*}
f_{j}(x) \leqq f_{j}(u) \quad \text { for all } j \in P=\{1,2, \ldots, p\} \tag{29}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{j}(x)<f_{i}(u) \quad \text { for some } \quad i \in P \tag{30}
\end{equation*}
$$

Proof. For each feasible $x$ for (VOP) and each feasible $(u, \tau, y$ ) for (DVOP) we have $y^{t} g(x)-y^{\prime} g(u) \leqq 0$; and since $y^{t} g(\cdot)$ is quasiconvex at $u$ this implies

$$
\begin{equation*}
(x-y)^{\prime} \nabla y^{\prime} g(u) \leqq 0 \tag{31}
\end{equation*}
$$

Applying (31) to $\sum_{i=1}^{p} \tau_{i} \nabla f_{i}(u)+\nabla y^{t} g(u)=0$ yields

$$
\begin{equation*}
(x-u)^{t} \sum_{i=1}^{p} \tau_{i} \nabla f_{i}(u) \geqq 0 \tag{32}
\end{equation*}
$$

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

$$
\begin{equation*}
\tau_{j} f_{j}(x) \leqq \tau_{j} f_{j}(u) \quad \text { for all } \quad j \in P=\{1,2, \ldots, p\} \tag{33}
\end{equation*}
$$

and

$$
\begin{equation*}
\tau_{i} f_{i}(x)<\tau_{i} f_{i}(u) \quad \text { for some } \quad i \in P \tag{34}
\end{equation*}
$$

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

$$
\begin{equation*}
\sum_{i=1}^{p} \tau_{i} f_{i}(x)<\sum_{i=1}^{p} \tau_{i} f_{i}(u) \tag{35}
\end{equation*}
$$

By hypothesis (a), i.e., $f_{i}$ 's are pseudoconvex, (33) and (34) imply

$$
\begin{equation*}
(x-u)^{t}\left(\sum_{i=1}^{p} \tau_{i} \nabla f_{i}(u)\right)<0 \tag{36}
\end{equation*}
$$

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 $\tau_{i} \geqq 0, i=1,2, \ldots, p$, (29), and (30) we obtain

$$
\begin{equation*}
\sum_{i=1}^{p} \tau_{i} f_{i}(x) \leqq \sum_{i=1}^{p} \tau_{i} f_{i}(u) \tag{37}
\end{equation*}
$$

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 $(u, \tau, y)$ for ( $D V O P$ ), $f_{i}, i=1,2, \ldots, p$, are $\rho_{i}$-convex and $g_{j}$, $j=1,2, \ldots, m$, are $\sigma_{j}$-convex. If either
(a) $\tau_{i}>0$ for all $i \in P=\{1,2, \ldots, p\}$ and $\sum_{i=1}^{p} \tau_{i} \rho_{i}+\sum_{j=1}^{m} y_{j} \sigma_{j} \geqq 0$, or
(b) $\sum_{i=1}^{p} \tau_{i} \rho_{i}+\sum_{j=1}^{m} y_{j} \sigma_{j}>0$,
then the following cannot hold:

$$
\begin{equation*}
f_{i}(x)<f_{i}(u) \quad \text { for some } \quad i \in P=\{1,2, \ldots, p\} \tag{38}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{j}(x) \leqq f_{j}(u) \quad \text { for all } \quad j \in P . \tag{39}
\end{equation*}
$$

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

$$
\begin{equation*}
\sum_{i=1}^{p} \tau_{i} f_{i}(x)<\sum_{i=1}^{p} \tau_{i} f_{i}(u) ; \tag{40}
\end{equation*}
$$

and for $\tau_{i} \geqq 0, i=1,2, \ldots, p$, (38) and (39) imply

$$
\begin{equation*}
\sum_{i=1}^{p} \tau_{i} f_{i}(x) \leqq \sum_{i=1}^{p} \tau_{i} f_{i}(u) . \tag{41}
\end{equation*}
$$

Now since $f_{i}$ 's are $\rho_{i}$-convex (40) and (41) imply

$$
\begin{gather*}
(x-u)^{t}\left(\sum_{i=1}^{p} \tau_{i} \nabla f_{i}(u)\right)+\left(\sum_{i=1}^{p} \tau_{i} \rho_{i}\right)\|x-u\|^{2}<0,  \tag{42}\\
(x-u)^{t}\left(\sum_{i=1}^{p} \tau_{i} \nabla f_{i}(u)\right)+\left(\sum_{i=1}^{p} \tau_{i} \rho_{i}\right)\|x-u\|^{2} \leqq 0, \tag{43}
\end{gather*}
$$

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

$$
\begin{equation*}
y^{\prime} g(x)-y^{\prime} g(u) \leqq 0 ; \tag{44}
\end{equation*}
$$

and because $g$ 's are $\sigma_{j}$-convex (44) implies

$$
\begin{equation*}
(x-u)^{t} \nabla y^{t} g(u)+\|x-u\|^{2} \sum_{j=1}^{m} y_{j} \sigma_{j} \leqq 0 . \tag{45}
\end{equation*}
$$

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

$$
\sum_{i=1}^{p} \tau_{i} \rho_{i}+\sum_{j=1}^{m} y_{j} \sigma_{j} \geqq 0,
$$

yield

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

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

Corollary 2. Assume weak duality (Theorem 4 or 5) holds between (VOP) and (DVOP). If $\left(u^{0}, \tau^{0}, y^{0}\right)$ is feasible for (DVOP) such that $u^{0}$ is feasible for (VOP), then $u^{0}$ is efficient for (VOP) and $\left(u^{0}, \tau^{0}, y^{0}\right)$ 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 therc exists a feasible $x$ for (VOP) such that (29) and (30) or (38) and (39) hold. But ( $u^{0}, \tau^{0}, y^{0}$ ) 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 ( $u^{0}, \tau^{0}, y^{0}$ ) is not efficient for (DVOP) leads to a contradiction, and hence $\left(u^{0}, \tau^{0}, y^{0}\right)$ 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 $P_{k}\left(\varepsilon^{0}\right)$ for at least one $k=1,2, \ldots, p$. Then there exist $\tau^{0} \in \mathbb{R}^{p}$ and $y^{0} \in \mathbb{R}^{m}$ such that $\left(x^{0}, \tau^{0}, y^{0}\right)$ is feasible for (DVOP). If also weak duality (Theorem 4 or 5) holds between (VOP) and (DVOP) then ( $x^{0}, \tau^{0}, y^{0}$ ) is efficient for (DVOP).
Proof. Since $x^{0}$ is an efficient solution of (VOP), then from Lemma 2, $x^{0}$ solves $P_{k}\left(\varepsilon^{0}\right)$ for each $k=1,2, \ldots, p$. By hypothesis there exists at least one $k=1,2, \ldots, p$ such that $x^{0}$ satisfies a constraint qualification [4,5] for $P_{k}\left(\varepsilon^{0}\right)$. From Kuhn-Tucker necessary conditions [4,5] we obtain $\tau_{i} \geqq 0$ for all $i \neq k$, and $y \geqq 0 \in \mathbb{R}^{m}$ such that

$$
\begin{align*}
\nabla f_{k}\left(x^{0}\right)+\sum_{i \neq k} \tau_{i} \nabla f_{i}\left(x^{0}\right)+\nabla y^{t} g\left(x^{0}\right) & =0  \tag{47}\\
y^{t} g\left(x^{0}\right) & =0 . \tag{48}
\end{align*}
$$

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

$$
\tau_{k}^{0}=\frac{1}{1+\sum_{i \neq k} \tau_{i}}>0, \quad \tau_{j}^{0}=\frac{\tau_{j}}{1+\sum_{i \neq k} \tau_{i}} \geqq 0 \quad \text { for all } j \neq k
$$

and

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

we conclude that ( $x^{0}, \tau^{0}, y^{0}$ ) is feasible for (DVOP). The efficiency of ( $x^{0}, \tau^{0}, y^{0}$ ) for (DVOP) now follows from Corollary 2.

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