

DUALITY IN NONCONVEX VECTOR MINIMUM PROBLEMS

T.R. GULATI AND NADIA TALAAT

A nonlinear vector minimum problem is considered. Duality theorems are proved for Mond-Weir type dual and their application to a certain nonlinear fractional vector minimum problem is discussed.

1. INTRODUCTION

Consider the vector minimum problem:

$$(P) \quad \text{minimise } f(x) \text{ subject to } g(x) \leq 0$$

where $f: R^n \rightarrow R^k$ and $g: R^n \rightarrow R^m$ are differentiable functions. Wolfe and Mond-Weir type duality for (P) has been discussed in several papers [1, 7, 8, 9, 10] using different concepts of optimality, namely: weak efficient (or weak minimum), efficient (or nondominated or noninferior or Pareto optimal) and properly efficient solutions.

Weir [8] proved weak and strong duality theorems for Mond-Weir [5] type dual of (P). In the strong duality theorem he obtained an efficient solution of the dual from a properly efficient solution of the primal problem. Bector *et al* [1] also discussed a similar result under stronger convexity assumptions. The duality results in Singh [7] and Weir and Mond [10] are for efficient and weak efficient solutions respectively. The converse duality theorems in [1, 7] are proved using Kuhn-Tucker type necessary conditions of Singh [6] and therefore need a constraint qualification.

In the present paper we also discuss duality results for Mond-Weir type dual of (P). Our results are different than those in [1, 7, 8, 10]. The strong duality theorem provides a properly efficient solution of the dual while in the converse duality theorem a weak efficient solution of the dual gives a properly efficient solution of the primal problem. Moreover, the converse duality theorem is proved using Fritz John necessary conditions [2] which do not need a constraint qualification.

Received 22 December 1990

The second author would like to thank the Government of India for providing her financial support under Indo-ARE Cultural Exchange Programme.

Copyright Clearance Centre, Inc. Serial-fee code: 0004-9729/91 \$A2.00+0.00.

2. PRELIMINARIES

The following convention of vectors in R^n will be followed throughout this paper: $x \leq y \Leftrightarrow x_i \leq y_i, i = 1, 2, \dots, n; x \leq y \Leftrightarrow x \leq y, x \neq y; x < y \Leftrightarrow x_i < y_i, i = 1, 2, \dots, n.$ $\nabla g(x)$ will denote the $m \times n$ Jacobian matrix of g at x , the index set $K = \{1, 2, \dots, k\}$ and $K_i = K - \{i\}$. For other notations and definitions we refer to Mangasarian [4].

Geoffrion [3] introduced the following scalar parametric problem:

$$(P_\lambda) \quad \text{minimise } \lambda^T f(x) \text{ subject to } g(x) \leq 0$$

and related its optimal solution to a properly efficient solution of (P) as follows:

LEMMA 1. *Let $\lambda > 0$ be fixed. If \bar{x} is an optimal solution of (P_λ) , then \bar{x} is a properly efficient solution of (P) .*

The Comprehensive Theorem in Geoffrion [3] includes the following necessary and sufficient conditions for problem (P) :

LEMMA 2. (Kuhn-Tucker type necessary conditions). *Let \bar{x} be a properly efficient solution of problem (P) and let g satisfy the Kuhn-Tucker constraint qualification at \bar{x} . Then there exist $\bar{\lambda} \in R^k, \bar{\mu} \in R^m$ such that*

$$\begin{aligned} \bar{\lambda}^T \nabla f(\bar{x}) + \bar{\mu}^T \nabla g(\bar{x}) &= 0 \\ \bar{\mu}^T g(\bar{x}) &= 0 \\ \bar{\lambda} > 0, \bar{\mu} &\geq 0. \end{aligned}$$

LEMMA 3. (Kuhn-Tucker type sufficient conditions). *Let f and g be convex. If there exist $\bar{\lambda} \in R^k$ and $\bar{\mu} \in R^m$ such that*

$$\begin{aligned} \bar{\lambda}^T \nabla f(\bar{x}) + \bar{\mu}^T \nabla g(\bar{x}) &= 0 \\ \bar{\mu}^T g(\bar{x}) &= 0 \\ \bar{\lambda} > 0, \bar{\mu} &\geq 0, \end{aligned}$$

then \bar{x} is a properly efficient solution of (P) .

3. DUALITY

In relation to (P) we consider the following Mond-Weir type dual:

$$\begin{aligned} (D) \quad & \text{maximise } f(y) \\ (1) \quad & \text{subject to } \lambda^T \nabla f(y) + \mu^T \nabla g(y) = 0 \\ (2) \quad & \mu^T g(y) \geq 0 \\ (3) \quad & \lambda > 0, \mu \geq 0. \end{aligned}$$

THEOREM 3.1. (Weak Duality). *Let x be feasible for (P) and (y, λ, μ) be feasible for (D) . If $\lambda^T f$ is pseudoconvex and $\mu^T g$ is quasiconvex at y , then*

$$(4) \quad \lambda^T f(x) \geq \lambda^T f(y).$$

PROOF: Since $g(x) \leq 0$ and $\mu \geq 0$,

$$\mu^T g(x) \leq 0 \leq \mu^T g(y).$$

Using quasiconvexity of $\mu^T g$ at y , we get

$$\mu^T \nabla g(y)(x - y) \leq 0.$$

Therefore from equation (1)

$$\lambda^T \nabla f(y)(x - y) \geq 0.$$

But $\lambda^T f$ is pseudoconvex at y . Hence

$$\lambda^T f(x) \geq \lambda^T f(y).$$

□

REMARK 1. It may be noted that the inequality (4) implies $f(x) \not\leq f(y)$.

THEOREM 3.2. *Let \bar{x} be feasible for (P) and $(\bar{y}, \bar{\lambda}, \bar{\mu})$ be feasible for (D) such that*

$$(5) \quad \bar{\lambda}^T f(\bar{x}) = \bar{\lambda}^T f(\bar{y}).$$

If $\bar{\lambda}^T f$ is pseudoconvex and $\bar{\mu}^T g$ is quasiconvex at \bar{y} , then \bar{x} is properly efficient for (P) .

PROOF: Let x be any feasible solution for (P) . From the weak duality theorem,

$$\bar{\lambda}^T f(x) \geq \bar{\lambda}^T f(\bar{y}).$$

Using (5), we get

$$\bar{\lambda}^T f(x) \geq \bar{\lambda}^T f(\bar{x}).$$

Thus \bar{x} is optimal for $(P_{\bar{\lambda}})$. Hence by Lemma 1, \bar{x} is properly efficient for (P) . □

THEOREM 3.3. *Let \bar{x} be feasible for (P) and $(\bar{y}, \bar{\lambda}, \bar{\mu})$ be feasible for (D) such that*

$$(6) \quad f(\bar{x}) = f(\bar{y}).$$

If $\lambda^T f$ is pseudoconvex and $\mu^T g$ is quasiconvex at y for each dual feasible (y, λ, μ) , then $(\bar{y}, \bar{\lambda}, \bar{\mu})$ is properly efficient for the dual problem (D) .

PROOF: First we show that $(\bar{y}, \bar{\lambda}, \bar{\mu})$ is efficient for (D) . Assume that it is not efficient, then there exists (y^*, λ^*, μ^*) feasible for (D) such that

$$f_r(y^*) > f_r(\bar{y}) \text{ for some } r \in K$$

$$f_i(y^*) \geq f_i(\bar{y}) \text{ for all } i \in K_r.$$

Therefore

or using (6),

$$\lambda^{*T} f(y^*) > \lambda^{*T} f(\bar{y})$$

$$\lambda^{*T} f(y^*) > \lambda^{*T} f(\bar{x}),$$

a contradiction to the weak duality theorem. Hence $(\bar{y}, \bar{\lambda}, \bar{\mu})$ is efficient for (D) . Assume now that it is not properly efficient. Then there exists a dual feasible solution (y^*, λ^*, μ^*) and an $i \in K$ such that $f_i(y^*) > f_i(\bar{y})$ and

$$f_i(y^*) - f_i(\bar{y}) > M(f_j(\bar{y}) - f_j(y^*))$$

for all $M > 0$ and all $j \in K_i$ satisfying $f_j(\bar{y}) > f_j(y^*)$. This means that $f_i(y^*) - f_i(\bar{y})$ can be made arbitrary large whereas $f_j(\bar{y}) - f_j(y^*)$ is finite for all $j \in K_i$. Therefore

$$\lambda_i^*(f_i(y^*) - f_i(\bar{y})) > \sum_{j \in K_i} \lambda_j^*(f_j(\bar{y}) - f_j(y^*))$$

or

$$\lambda^{*T} f(y^*) > \lambda^{*T} f(\bar{y}).$$

Using (6), we get

$$\lambda^{*T} f(y^*) > \lambda^{*T} f(\bar{x}).$$

This again contradicts the weak duality theorem. Hence $(\bar{y}, \bar{\lambda}, \bar{\mu})$ is a properly efficient solution for (D) . □

THEOREM 3.4. (Strong Duality). *Let \bar{x} be a properly efficient solution of Problem (P) and let g satisfy the Kuhn-Tucker constraint qualification at \bar{x} . Then there exists $(\bar{\lambda}, \bar{\mu})$, such that $(\bar{y} = \bar{x}, \bar{\lambda}, \bar{\mu})$ is a feasible solution for (D) and the objective values of (P) and (D) are equal. Also, if $\lambda^T f$ is pseudoconvex and $\mu^T g$ is quasiconvex at y for every dual feasible solution (y, λ, μ) , then $(\bar{x}, \bar{\lambda}, \bar{\mu})$ is a properly efficient solution for (D) .*

PROOF: Since \bar{x} is a properly efficient solution for (P) at which the Kuhn-Tucker constraint qualification is satisfied, by Lemma 2 there exist $\bar{\lambda} \in R^k$ and $\bar{\mu} \in R^m$ such that

$$\bar{\lambda}^T \nabla f(\bar{x}) + \bar{\mu}^T \nabla g(\bar{x}) = 0$$

$$\bar{\mu}^T g(\bar{x}) = 0$$

$$\bar{\lambda} > 0, \quad \bar{\mu} \geq 0.$$

Therefore $(\bar{x}, \bar{\lambda}, \bar{\mu})$ is feasible for (D) . Also, the two vector objectives are equal. Hence, by Theorem 3.3, $(\bar{x}, \bar{\lambda}, \bar{\mu})$ is a properly efficient solution for (D) . \square

THEOREM 3.5. (Converse Duality). *Let $(\bar{y}, \bar{\lambda}, \bar{\mu})$ be a weak efficient solution for (D) , the $n \times n$ Hessian matrix $\nabla^2(\bar{\lambda}^T f(\bar{y}) + \bar{\mu}^T g(\bar{y}))$ be positive or negative definite and $\nabla f_i(\bar{y}), i = 1, 2, \dots, k$ be linearly independent. If $\lambda^T f$ is pseudoconvex and $\mu^T g$ is quasiconvex at \bar{y} , then \bar{y} is a properly efficient solution for (P) .*

PROOF: Since $(\bar{y}, \bar{\lambda}, \bar{\mu})$ is a weak efficient solution for (D) , by Theorem 1 [2] there exist $\bar{u} \in R^k, \bar{v} \in R^n, \bar{w} \in R, \bar{\eta} \in R^k$ and $\bar{\nu} \in R^m$ such that

$$\begin{aligned}
 (7) \quad & \bar{u}^T \nabla f(\bar{y}) + \bar{v}^T \nabla^2(\bar{\lambda}^T f(\bar{y}) + \bar{\mu}^T g(\bar{y})) + \bar{w} \bar{\mu}^T \nabla g(\bar{y}) = 0 \\
 (8) \quad & [\nabla f(\bar{y})] \bar{v} + \bar{\eta} = 0 \\
 (9) \quad & [\nabla g(\bar{y})] \bar{v} + \bar{w} g(\bar{y}) + \bar{\nu} = 0 \\
 (10) \quad & \bar{w} [\bar{\mu}^T g(\bar{y})] = 0 \\
 (11) \quad & \bar{\eta}^T \bar{\lambda} = 0 \\
 (12) \quad & \bar{\nu}^T \bar{\mu} = 0 \\
 (13) \quad & (\bar{u}, \bar{w}, \bar{\eta}, \bar{\nu}) \geq 0, \quad (\bar{u}, \bar{v}, \bar{w}, \bar{\eta}, \bar{\nu}) \neq 0.
 \end{aligned}$$

Now $\bar{\lambda} > 0, \bar{\eta}^T \bar{\lambda} = 0 \Rightarrow \bar{\eta} = 0$. Therefore from (8)

$$(14) \quad [\nabla f(\bar{y})] \bar{v} = 0$$

which with equation (1) gives

$$(15) \quad \bar{\mu}^T \nabla g(\bar{y}) \bar{v} = 0.$$

On multiplying (7) by \bar{v} from the right and using equations (14) and (15), we get

$$(16) \quad \bar{v}^T [\nabla^2(\bar{\lambda}^T f(\bar{y}) + \bar{\mu}^T g(\bar{y}))] \bar{v} = 0.$$

Since $\nabla^2(\bar{\lambda}^T f(\bar{y}) + \bar{\mu}^T g(\bar{y}))$ is assumed to be positive or negative definite, equation (16) gives $\bar{v} = 0$.

Now suppose $\bar{w} = 0$. Therefore from equation (9), $\bar{\nu} = 0$. Also, equation (7) gives

$$(17) \quad \bar{u}^T \nabla f(\bar{y}) = 0.$$

Since $\nabla f_i(\bar{y})$ are assumed to be linearly independent, (17) implies $\bar{u} = 0$. Thus we get

$$(\bar{u}, \bar{v}, \bar{w}, \bar{\eta}, \bar{\nu}) = 0,$$

a contradiction to (13). Hence $\bar{w} > 0$.

Now from equation (9),

$$g(\bar{y}) = \frac{-\bar{v}}{\bar{w}} \leq 0.$$

Therefore \bar{y} is a feasible solution for (P), and Theorem 3.2 implies that \bar{y} is a properly efficient solution for (P).

The following result gives sufficient conditions for a weak efficient solution of the dual problem (D) to be properly efficient. It follows immediately from Theorems 3.3 and 3.5. □

COROLLARY. *Let $(\bar{y}, \bar{\lambda}, \bar{\mu})$ be a weak efficient solution for (D) and let the hypotheses of Theorem 3.5 be satisfied. If $\lambda^T f$ is pseudoconvex and $\mu^T g$ is quasiconvex at y for each dual feasible (y, λ, μ) , then $(\bar{y}, \bar{\lambda}, \bar{\mu})$ is a properly efficient solution of the dual problem (D).*

REMARK 2. It may be noted that in Weir and Mond [10] the dual variable $\lambda \geq 0$. Therefore, their implication $w^T \lambda_0 = 0 \Rightarrow w = 0$ in the converse duality theorem is erroneous.

THEOREM 3.6. (Strict Converse Duality Theorem). *Let \bar{x} and $(\bar{y}, \bar{\lambda}, \bar{\mu})$ be feasible solutions for (P) and (D) respectively such that*

$$(18) \quad \bar{\lambda}^T f(\bar{x}) = \bar{\lambda}^T f(\bar{y}).$$

If $\bar{\lambda}^T f$ is strictly pseudoconvex and $\bar{\mu}^T g$ is quasiconvex at \bar{y} , then $\bar{x} = \bar{y}$ and \bar{y} is properly efficient for (P).

PROOF: We will assume $\bar{x} \neq \bar{y}$ and exhibit a contradiction.

Since \bar{x} and $(\bar{y}, \bar{\lambda}, \bar{\mu})$ are feasible for (P) and (D) respectively, we have

$$\bar{\mu}^T g(\bar{x}) \leq 0 \leq \bar{\mu}^T g(\bar{y}).$$

The quasiconvexity of $\bar{\mu}^T g$ at \bar{y} implies

$$(19) \quad \bar{\mu}^T \nabla g(\bar{y})(\bar{x} - \bar{y}) \leq 0.$$

Equations (1) and (19) imply

$$\bar{\lambda}^T \nabla f(\bar{y})(\bar{x} - \bar{y}) \geq 0.$$

But $\bar{\lambda}^T f$ is strictly pseudoconvex at \bar{y} . Hence

$$\bar{\lambda}^T f(\bar{x}) > \bar{\lambda}^T f(\bar{y}),$$

a contradiction to (18). Hence $\bar{x} = \bar{y}$. Proper efficiency of \bar{y} for (P) now follows from Theorem 3.2. □

4. APPLICATION

We now apply our results to obtain a dual for the following nonlinear fractional vector minimum problem:

$$(FP) \quad \text{minimise } f(x) = \left[\frac{\phi_1(x)}{\psi(x)}, \frac{\phi_2(x)}{\psi(x)}, \dots, \frac{\phi_k(x)}{\psi(x)} \right]$$

$$\text{subject to } g(x) \leq 0, \quad x \in S$$

where

- (i) $S \subseteq R^n$ is an open convex set,
- (ii) $\phi: S \rightarrow R^k, g: S \rightarrow R^m$ are differentiable convex functions on S and $\psi: S \rightarrow R$ is a differentiable concave function on S , and
- (iii) $\phi(x) \geq 0$ and $\psi(x) > 0$ on S .

Therefore for each $\lambda > 0, \lambda^T f$ is pseudoconvex and since g is convex, convexity hypotheses in this paper are satisfied. Hence the dual problem (D) becomes

$$\text{maximise } f(y) = \left[\frac{\phi_1(y)}{\psi(y)}, \frac{\phi_2(y)}{\psi(y)}, \dots, \frac{\phi_k(y)}{\psi(y)} \right]$$

$$\text{subject to } \sum_{i=1}^k \lambda_i \left[\frac{\nabla \phi_i(y)}{\psi(y)} - \frac{\phi_i(y)}{\psi^2(y)} \nabla \psi(y) \right] + \mu^T \nabla g(y) = 0$$

$$\mu^T g(y) \geq 0, \quad \lambda > 0, \mu \geq 0, y \in S.$$

On simplification, we get the following dual for (FP):

$$(FD) \quad \text{maximise } f(y)$$

$$\text{subject to}$$

$$\lambda^T \nabla \phi(y) - \left(\sum_{i=1}^k v_i \right) \nabla \psi(y) + w^T \nabla g(y) = 0$$

$$w^T g(y) \geq 0$$

$$\lambda_i \phi_i(y) - v_i \psi(y) = 0, \quad i = 1, 2, \dots, k$$

$$\lambda > 0, v \geq 0, w \geq 0, y \in S$$

where $\lambda, v \in R^k$ and $w \in R^m$.

In particular when $k = 1$, we get the following pair of scalar nonlinear program and its dual

Primal Problem:

$$\begin{aligned} & \text{minimise } \frac{\phi(x)}{\psi(x)} \\ & \text{subject to } g(x) \geq 0, \quad x \in S \end{aligned}$$

Dual Problem:

$$\begin{aligned} & \text{maximise } f(y) = \frac{\phi(y)}{\psi(y)} \\ & \text{subject to} \\ & \quad \nabla\phi(y) - v\nabla\psi(y) + w^T\nabla g(y) = 0 \\ & \quad \quad \quad w^T g(y) \geq 0 \\ & \quad \quad \quad \phi(y) - v\psi(y) = 0 \\ & \quad \quad \quad v, w \geq 0, y \in S \end{aligned}$$

where $\phi: S \rightarrow R$ and $v \in R$. Other notations are as in (FD).

REMARK 3. If ψ is linear, then $\phi(x) \geq 0$ is not required and the dual variable v in the above dual problems is unrestricted in sign.

REFERENCES

- [1] C.R. Bector, S. Chandra and M.V. Durgaprasad, 'Duality in pseudolinear multiobjective programming', *Asia-Pacific J. Oper. Res.* **5** (1988), 150-159.
- [2] B.D. Craven, 'Lagrangian conditions and quasiduality', *Bull. Austral. Math. Soc.* **16** (1977), 325-339.
- [3] A.M. Geoffrion, 'Proper efficiency and the theory of vector maximization', *J. Math. Anal. Appl.* **22** (1968), 618-630.
- [4] O.L. Mangasarian, *Nonlinear programming* (McGraw-Hill, New York, 1969).
- [5] B. Mond and T. Weir, 'Generalized concavity and duality', in *Generalized concavity in optimization and economics*, Editors S. Schaible and W.T. Ziemba, pp. 263-279 (Academic Press, New York, 1981).
- [6] C. Singh, 'Optimality conditions in multiobjective differentiable programming', *J. Optim. Theory Appl.* **53** (1987), 115-123.
- [7] C. Singh, 'Duality theory in multiobjective differentiable programming', *J. Inform. Optim. Sci.* **9** (1988), No. 2, 231-240.
- [8] T. Weir, 'Proper efficiency and duality for vector valued optimization problems', *J. Austral. Math. Soc., Series A* **43** (1987), 21-34.
- [9] T. Weir and V. Jeyakumar, 'A class of nonconvex functions and mathematical programming', *Bull. Austral. Math. Soc.* **38** (1988), 177-189.

- [10] T. Weir and B. Mond, 'Generalized convexity and duality in multiple objective programming', *Bull. Austral. Math. Soc.* **39** (1989), 287–299.

Department of Mathematics
University of Roorkee
Roorkee
India 247667

Department of Mathematics
University of Alexandria
Egypt