

Fritz John Type Optimality Criteria for Optimal Solution of Nonlinear Programming Problem

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Abstract: This paper is proposed to study Fritz-John optimality and duality results for nonlinear programming problem. Weir and Mond provided a Fritz John dual for the nonlinear programming problems involving differentiable functions by using the Fritz John optimality conditions instead of Kuhn-Tucker conditions and thus did not require a constraint qualification. Here we associate the Mond Wier type Fritz John dual to the nonlinear programming problem.

1. INTRODUCTION

Weir and Mond [7] provided a Fritz John dual for the nonlinear programming problems involving differentiable functions by using the Fritz John [4] optimality conditions instead of Kuhn-Tucker [5] conditions and thus did not require a constraint qualification.

Keywords: Nonlinear programming, Convex functions, Fritz-John optimality and duality conditions, Kuhn-Tucker conditions.

1. Introduction

We obtain Fritz John type necessary optimality criteria for the optimal solution of the following nonlinear program:

(P) Min $f(x)$

Subject to $g_j \leq 0, j = 1, 2, 3, \dots, m, x \in S,$

Where $S \subseteq R^n, f: S \rightarrow R,$ and

$g_j: S \rightarrow R, j = 1, 2, 3, \dots, m$ are real valued functions. S

is a locally connected set such that for each $x^*, x \in S,$ there exists a vector value function $H_{x^*, x(\lambda)},$ satisfying:

$$H_{x^*, x(\lambda)} \in S, 0 < \lambda < a \quad (1.1)$$

Let $X^0 = \{x \in S / g_j \leq 0, j = 1, 2, 3, \dots, m\}$ $H_{x^*, x}$ is

continuous in the interval $]0, a(x^*, x)[$ and

$$H_{x^*, x(0)} = x^*, H_{x^*, x(1)} = x \quad (1.2)$$

And the right differentials of f and $g_j, j = 1, 2, 3, \dots, m$ at x^* exist with respect to the arc $H_{x^*, x(\lambda)}.$

Theorem-1 Let x^* be an optimal solution of (P). If $(df)^+(x^*, H_{x^*, x(0+)})$ and $(dg_I)^+(x^*, H_{x^*, x(0+)})$ are convex functions of x and $g_j, j \in J$ is continuous at x^* with S convex or $S = R^n$ then there exist $r_0^* \in R, r^* \in R,$ Such that

$$r_0^* (df)^+(x^*, H_{x^*, x(0+)}) + (r_0^*)^T (dg_I)^+(x^*, H_{x^*, x(0+)}) \geq 0, \text{ for all } x \in S \quad (1.3)$$

$$(r_0^*)^T g(x^*) = 0 \quad (1.4)$$

$$(r_0^*, r^*) \geq 0 \quad (1.5)$$

Where $I = I(x^*) = \{i / g_i(x^*) = 0\}$

and $J = J(x^*) = \{j / g_j(x^*) < 0\}.$

Proof: First we shall show that the system

$$\begin{cases} (df)^+(x^*, H_{x^*, x(0+)}) < 0 \\ (dg_I)^+(x^*, H_{x^*, x(0+)}) < 0 \end{cases} \quad (1.6)$$

Has no solution in $x \in S.$

If possible let $x \in S$ be a solution of the system (1.6). Since right differentials of f and $g_j, i \in I$ at x^* exist with respect to the arc $H_{x^*, x(\lambda)}.$ Therefore

$$f(H_{x^*, x(\lambda)}) = f(x^*) + \lambda(df)^+(x^*, H_{x^*, x(0+)}) + \lambda\alpha(\lambda) \quad (1.7)$$

$$\text{and } g_i(H_{x^*,x^{(\lambda)}}) = g_i(x^*) + \lambda(dg_i)^+(x^*, H_{x^*,x^{(0+)}}) + \lambda\alpha_i(\lambda) \quad (1.8)$$

$$\text{where } \alpha: [0,1] \rightarrow R, \lim_{\lambda \rightarrow 0} \alpha(\lambda) = 0 \quad (1.9)$$

$$\alpha_i: [0,1] \rightarrow R, \lim_{\lambda \rightarrow 0^+} \alpha_i(\lambda) = 0, i \in I(x^*) \quad (1.10)$$

Using (1.6),(1.9) and (1.10) we get for small enough λ , say $0 < \lambda < \lambda_0$ then $(df)^+(x^*, H_{x^*,x^{(0+)}}) + \alpha(\lambda) < 0$ and

$$(dg_i)^+(x^*, H_{x^*,x^{(0+)}}) + \alpha_i(\lambda) < 0, \quad i \in I(x^*)$$

hence it follows by using the relation(1.7) and (1.8) that for $0 < \lambda < \lambda_0$,

$$f(H_{x^*,x^{(\lambda)}}) - f(x^*) < 0 \quad (1.11)$$

$$g_i(H_{x^*,x^{(\lambda)}}) - g_i(x^*) < 0, i \in I(x^*) \quad (1.12)$$

Now $g_j, j \in J$ is continuous at x^* and $H_{x^*,x^{(\lambda)}}$ is also a continuous function of λ . Therefore

$$\lim_{\lambda \rightarrow 0^+} g_j(H_{x^*,x^{(\lambda)}}) = g_j(x^*) < 0$$

Which implies that there exist λ_j^* ,

$$0 < \lambda_j^* < a(x^*, x), j \in J \text{ such that}$$

$$g_j(H_{x^*,x^{(\lambda)}}) < 0 \text{ for } 0 < \lambda < \lambda_j^* \quad (1.13)$$

Let $\lambda^* = \min(\lambda_0, \lambda_j^*, j \in J)$ then from (1.11) to (1.13) it follows that for $0 < \lambda < \lambda_j^*$, $H_{x^*,x^{(\lambda)}} \in X^0$ and $f(H_{x^*,x^{(\lambda)}}) < f(x^*)$, which is a contradiction as x^* is an optimal solution of (P).

Hence the system (1.6) has no solution $x \in S$.

Since $(df)^+(x^*, H_{x^*,x^{(0+)}})$ and $(dg_i)^+(x^*, H_{x^*,x^{(0+)}})$, $i \in I(x^*)$ are convex functions of x therefore there exist $r_0^* \in R, r_i^* \in R, i \in I$ such that

$$r_0^*(df)^+(x^*, H_{x^*,x^{(0+)}}) + r_i^{*T}(dg_i)^+(x^*, H_{x^*,x^{(0+)}}) \geq 0 \text{ for all } x \in S, \text{ where } (r_0^*, r_i^*) \geq 0.$$

Defining $r_j^* = 0$ we get the required result.

Now we associate the following Mond Wier type Fritz John dual to the problem (P):

$$(D) \text{ Maximize } f(\mu)$$

Subject to

$$r_0(df)^+(\mu, H_{\mu,x^{(0+)}}) + r^T(dg)^+(\mu, H_{\mu,x^{(0+)}}) \geq 0 \text{ for all } x \in X^0 \quad (1.14)$$

$$\sum_{j=1}^m r_j g_j(\mu) \geq 0 \quad (1.15)$$

$$\mu \in S, (r_0, r) \geq 0, r_0 \in R, r \in R^m \quad (1.16)$$

Theorem-2 (Weak duality) Let x be feasible for (P) and (μ, r_0, r) be feasible for (D). If f is locally P-connected and

$\sum_{j=1}^m r_j g_j(\mu)$ is strongly P-connected at μ then $f(x) \geq f(\mu)$.

Proof: If possible let $f(x) < f(\mu)$. Since f is locally P-connected at μ therefore it follows that

$$r_0(df)^+(\mu, H_{\mu,x^{(0+)}}) \leq 0 \quad (1.17)$$

With strict inequality if $r_0 > 0$. by the feasibility of x and (μ, r_0, r) for (P) and (D) respectively we get

$$\sum_{j=1}^m r_j g_j(x) \leq \sum_{j=1}^m r_j g_j(\mu),$$

Now $\sum_{j=1}^m r_j g_j$ is strongly locally P-connected at μ so,

$$d(\sum_{j=1}^m r_j g_j)^+(\mu, H_{\mu,x^{(0+)}}) \leq 0 \quad (1.18)$$

With strict inequality if some $r_j > 0, j = 1, 2, \dots, m$. Adding (1.17), (1.18) and using (1.16) we get

$$r_0(df)^+(\mu, H_{\mu,x^{(0+)}}) + r^T(dg)^+(\mu, H_{\mu,x^{(0+)}}) < 0$$

Which is a contradiction to (1.14).

Hence $f(x) \geq f(\mu)$.

Theorem-3: (Strong Duality) Let x^* be an optimal solution of (P), $(df)^+(x^*, H_{x^*,x^{(0+)}})$ and $(dg_i)^+(x^*, H_{x^*,x^{(0+)}})$ be

the convex functions of x and g_j $j \in J$ be continuous at x^* with S convex or $S = R^n$. Then there exist $r_0^* \in R$ and $r_0^* \in R^m$ such that (x^*, r_0^*, r^*) is feasible for (D) and the values of the objective functions of (P) and (D) are at x^* . Also if for each feasible (μ, r_0, r) for (D), f is locally P-connected and $\sum_{j=1}^m r_j g_j$ is strongly locally P-connected at μ then (x^*, r_0^*, r^*) is optimal for (D).

Proof: Since x^* is an optimal solution of (P) therefore by theorem -1, there exist $r_0^* \in R$ and $r_0^* \in R^m$ such that (x^*, r_0^*, r^*) is feasible for (D). Equality of objective functions for (P) and (D) follows trivially. Further if (x^*, r_0^*, r^*) is not optimal for (D) then there exists (μ, r_0, r) ,feasible for (D) such that $f(\mu) > f(x^*)$.which is a contradiction to weak duality.

2. CONCLUSION

In theorem-1, we obtained Fritz John type necessary

optimality condition for nonlinear programming problem (P). In theorem-2 and 3, we obtained Mond Wier type duality conditions for nonlinear program by considering a dual (D) of (P).

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