

REMARKS ON NONSMOOTH DYNAMIC VECTOR OPTIMIZATION PROBLEMS

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1. Introduction. This paper deals with vector optimization problems. By convention, throughout this paper we will use the following notations. For $y = (y_1, \dots, y_n)$, $z = (z_1, \dots, z_n) \in R^n$, we say that

- (i) $y \leq z$, if and only if $y_i \leq z_i$ for any $i \in \{1, \dots, n\}$,
- (ii) $y < z$ if and only if $y_i \leq z_i$ for any $i \in \{1, \dots, n\}$ with $y \neq z$,
- (ii) $y \ll z$ if and only if $y_i < z_i$ for any $i \in \{1, \dots, n\}$.

Recently, many papers have been devoted to optimality conditions for the vector-valued programming and optimal control problems under some smooth or convex assumptions (see [2], [6], [7], [9], [10]). In [11], we derived the Kuhn-Tucker type proper-efficiency conditions for vector optimal control problems in general case. In this paper we use analogous method to discuss weak-efficiency and efficiency conditions for the following problem,

$$(P) : \quad \begin{aligned} & \text{minimize : } \mathcal{F}(x, u) := (\mathcal{F}_1(x, u), \dots, \mathcal{F}_k(x, u)) \\ & \text{subject to : } \dot{x}(t) = \Phi(t, x(t), u(t)) \quad \text{a.e.}, \\ & \quad \quad \quad x(0) \in D, \quad u(t) \in U(t) \quad \text{a.e.}, \\ & \quad \quad \quad \mathcal{G}(x, u) := (\mathcal{G}_1(x, u), \dots, \mathcal{G}_l(x, u)) \leq 0 \end{aligned}$$

where

$$\begin{aligned} \mathcal{F}_i(x, u) &:= \int_0^1 F_i(t, x(t), u(t))dt + f_i(x(1)) \text{ for } i \in I := \{1, \dots, k\} \\ \mathcal{G}_j(x, u) &:= \int_0^1 G_j(t, x(t), u(t))dt + g_j(x(1)) \text{ for } j \in J := \{1, \dots, l\}; \end{aligned}$$

$x(\cdot) \in AC([0, 1], R^m)$ and $u(\cdot) \in M([0, 1], R^n)$; $F_i, G_j : [0, 1] \times R^m \times R^n \rightarrow R$, $f_i, g_j : R^m \rightarrow R$ for $i \in I, j \in J$ and $\Phi : [0, 1] \times R^m \times R^n \rightarrow R^m$ are given functions; D is a subset of R^m and $U(\cdot) : [0, 1] \rightarrow 2^{R^n}$ is a set-valued function. Here, $AC([0, 1], R^m)$ is the space of absolutely continuous functions on $[0, 1]$ with value in R^m , $M([0, 1], R^n)$ is the space of Lebesgue measurable functions on $[0, 1]$ with value in R^n .

For this optimal control problem (P), we say that (x, u) is an admissible process iff $F_i(\cdot, x(\cdot), u(\cdot))$ and $G_j(\cdot, x(\cdot), u(\cdot))$ are integrable for every $i \in I$ and $j \in J$, (x, u) satisfies state equation $\dot{x}(t) = \Phi(t, x(t), u(t))$ a.e. with $x(0) \in D, u(t) \in U(t)$ a.e. and $\mathcal{G}(x, u) \leq 0$. The first component of a process (x, u) is called a state and the second is called a control. We denote by Ω the set of all admissible processes of (P). The optimal solutions for (P) are defined in the following meaning.

Definition 1: $(x_*, u_*) \in \Omega$ is said to be

- (i) a weakly-efficient solution for (P) if there exists no $(x, u) \in \Omega$ such that

$$\mathcal{F}(x, u) \ll \mathcal{F}(x_*, u_*);$$

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(ii) an efficient solution for (P) if there exists no $(x, u) \in \Omega$ such that

$$\mathcal{F}(x, u) < \mathcal{F}(x_*, u_*).$$

Definition 2: $(x_*, u_*) \in \Omega$ is called a local weakly-efficient solution of type (I) (resp. (II)) for (P) if and only if there is no $(x, u) \in \Omega$ with $\|x - x_*\|_{L^\infty} \leq \epsilon$ for some $\epsilon > 0$ (resp. with $x(t) \in x_*(t) + \epsilon B_m$ and $u(t) \in u_*(t) + \epsilon B_n$ for some $\epsilon > 0$, where B^m and B^n are unit closed balls of R^m and R^n , respectively) such that $\mathcal{F}(x, u) \ll \mathcal{F}(x_*, u_*)$.

The main method to obtain optimality conditions for multiobjective optimization problems is based on a replacement of the multiobjective problems by single-objective (scalar) optimization problems. The following results give the relationship between (P) and scalar optimization problems.

Lemma 1: $(x_*, u_*) \in \Omega$ is a weakly-efficient (local weakly-efficient) solution of (P) if and only if (x_*, u_*) is an optimal (local optimal) solution of the following scalar optimization problem,

$$\begin{aligned} \min : & \max_{i \in I} (\mathcal{F}_i(x, u) - \mathcal{F}_i(x_*, u_*)) \\ \text{s. t. :} & (x, u) \in \Omega. \end{aligned}$$

Proof. By the definitions, it is easy to see that (x_*, u_*) is a weakly efficient of (P) if and only if there is no $(x, u) \in \Omega$ satisfying

$$\max_{i \in I} (\mathcal{F}_i(x, u) - \mathcal{F}_i(x_*, u_*)) < 0.$$

Thus, this lemma hold. \square

Lemma 2: ([6, Lemma 3.1]) $(x_*, u_*) \in \Omega$ is an efficient solution of (P) if and only if (x_*, u_*) is an optimal solution of the following scalar optimal control problem (P_i) for each $i \in I$.

$$\begin{aligned} (P_i) : & \text{minimize : } \mathcal{F}_i(x, u) \\ & \text{subject to : } (x, u) \in \Omega \\ & \mathcal{F}_j(x, u) - \mathcal{F}_j(x_*, u_*) \leq 0 \quad j \in I/\{i\}. \end{aligned}$$

Lemma 3: Suppose that Ω is convex set and $\mathcal{F}_i(x, u)$, $i = 1, \dots, k$ are convex functions. Then, $(x_*, u_*) \in \Omega$ is a weakly-efficient solution of (P) if and only if (x_*, u_*) is an optimal solution of (P_i) stated in Lemma 2 for some $i \in I$.

Proof. Assume that (x_*, u_*) is a weakly-efficient solution of (P). If for every (P_i) , (x_*, u_*) is not an optimal solution, i.e. for any $i \in I$ there exists $(x_i, u_i) \in \Omega$ with

$$\begin{aligned} \mathcal{F}_i(x_i, u_i) & < \mathcal{F}_i(x_*, u_*) \\ \mathcal{F}_j(x_i, u_i) - \mathcal{F}_j(x_*, u_*) & \leq 0 \quad \text{for } j \in I/\{i\}. \end{aligned}$$

Putting $(x_0, u_0) := \frac{1}{k} \sum_{i \in I} (x_i, u_i)$, we see that $(x_0, u_0) \in \Omega$. Notice that $\mathcal{F}_i(x, u)$ is convex, we have

$$\mathcal{F}_i(x_0, u_0) \leq \sum_{j \in I} \frac{1}{k} \mathcal{F}_i(x_j, u_j) < \mathcal{F}_i(x_*, u_*).$$

Thus, $\mathcal{F}(x_0, u_0) \ll \mathcal{F}(x_*, u_*)$, which contradicts that (x_*, u_*) is a weakly-efficient solution of (P) .

Conversely, let (x_*, u_*) be an optimal solution of (P_i) for some $i \in I$. If (x_*, u_*) is not a weakly-efficient solution of (P) , then there is $(x, u) \in \Omega$ satisfying

$$\mathcal{F}_i(x, u) < \mathcal{F}_i(x_*, u_*) \text{ and } \mathcal{F}_j(x, u) - \mathcal{F}_j(x_*, u_*) < 0 \text{ for } j \in I/\{i\},$$

which contradicts that (x_*, u_*) is an optimal solution of (P_i) . \square

2. Optimality conditions. For simplicity, throughout this section we omit the variable t when it does not cause confusion, and abbreviate the arguments $(t, x_*(t), u_*(t))$ to $[t]$, for instance, we write $G_i[t] = G_i(t, x_*(t), u_*(t))$. In Theorem 1 and 2 below, the notations ∂ denote the Clarke generalized gradients and $N_D, N_{U(t)}$ indicate the Clarke normal cones, while in Theorem 3 and 4, these notations stand for the subdifferentials and the normal cones in the sense of convex analysis, respectively.

The following assumptions are required. The pair (x_*, u_*) in (A2) and (A3) will be assumed to be a local weakly efficient solution of type (I) for (P) .

(A1): D is closed, $U(\cdot)$ is a nonempty compact set-valued map and the graph GrU is $\mathcal{L} \times \mathcal{B}$ measurable.

(A2): $f_i(\cdot), g_j(\cdot)$ ($i \in I, j \in J$) are Lipschitz continuous in a neighborhood of $x_*(1) \in R^m$.

(A3): For every admissible control $u(\cdot)$, there are real-valued measurable function $\epsilon(t) > 0$ and $h_i(t) \geq 0, i = 0, \dots, k+l$, such that

$$\begin{aligned} |F_i(t, x, u(t)) - F_i(t, x', u(t))| &\leq h_i(t) |x - x'| \text{ for } i \in I \\ |G_j(t, x, u(t)) - G_j(t, x', u(t))| &\leq h_{k+j}(t) |x - x'| \text{ for } j \in J \\ |\Phi(t, x, u(t)) - \Phi(t, x', u(t))| &\leq h_0(t) |x - x'| \end{aligned}$$

whenever $|x - x_*(t)| \leq \epsilon(t), |x' - x_*(t)| \leq \epsilon(t), t \in [0, 1]$; for $u(\cdot) = u_*(\cdot)$ these functions can be chosen in such a way that $\epsilon(t) = \epsilon > 0$ and $h_i(t)$ ($i = 0, \dots, k+l$) are integrable.

(A4): For any $u(\cdot) \in \mathcal{U} := \{u(\cdot) \in M([0, 1], R^n) : u(t) \in U(t) \text{ a.e.}\}$, $F_i(t, x, u(t))$ for $i \in I, G_j(t, x, u(t))$ for $j \in J$ and $\Phi(t, x, u(t))$ are measurable.

Theorem 1. *Let assumptions (A1)-(A4) be satisfied. Suppose that (x_*, u_*) is a local weakly efficient solution of type (I) for (P) . Then, there exist $\lambda = (\lambda_1, \dots, \lambda_{k+l}) > 0$ and an absolutely continuous function $p(\cdot) : [0, 1] \rightarrow R^n$, such that*

$$(1) \quad -\dot{p}(t) \in \partial_x H(t, x_*(t), p(t), u_*(t), \lambda) \quad \text{a.e.}$$

$$(2) \quad p(0) \in N_D(x_*(0)), \quad -p(1) \in \sum_{i \in I} \lambda_i \partial f_i(x_*(1)) + \sum_{j \in J} \lambda_{k+j} \partial g_j(x_*(1))$$

$$(3) \quad H(t, x_*(t), p(t), u_*(t), \lambda) = \max_{v \in U(t)} H(t, x_*(t), p(t), v, \lambda) \quad \text{a.e.}$$

$$(4) \quad \lambda_{k+j} \left(\int_0^1 G_j[t] dt + g_j(x_*(1)) \right) = 0 \quad \text{for } j \in J$$

where $H(t, x, p, u, \lambda) := \langle p, \Phi(t, x, u) \rangle - \sum_{i \in I} \lambda_i F_i(t, x, u) - \sum_{j \in J} \lambda_{k+j} G_j(t, x, u)$

Proof. We consider the following problem,

$$(P') \quad \begin{aligned} \min : \quad & \Gamma_0(y) := \max_{i \in I} \{y_i(1) + f_i(x(1)) - \mathcal{F}_i(x_*, u_*)\} \\ \text{s. t. :} \quad & L_0(y, u) := x(1) - x(0) - \int_0^1 \Phi(t, x(t), u(t)) dt = 0 \\ & L_i(y, u) := y_i(1) - \int_0^1 F_i(t, x(t), u(t)) dt = 0 \quad i \in I \\ & L_{k+j}(y, u) := y_{k+j}(1) - \int_0^1 G_j(t, x(t), u(t)) dt = 0 \quad j \in J \\ & \Gamma_j(y) := y_{k+j}(1) + g_j(x(1)) \leq 0 \quad j \in J \\ & y(\cdot) \in \mathcal{S}, \quad u(\cdot) \in \mathcal{U}, \end{aligned}$$

where $y(\cdot) := (x(\cdot), y_1(\cdot), \dots, y_{k+l}(\cdot)) \in C([0, 1], R^{m+k+l})$ is the state and $u(\cdot) \in M([0, 1], R^n)$ is the control, $\mathcal{S} := \{x \in C([0, 1], R^m) : x(0) \in D\} \times C([0, 1], R^{2k})$.

Let $y_{i_*}(t) := \int_0^t F_i[t] dt$ for $i \in I$ and $y_{(k+j)_*}(t) := \int_0^t G_j[t] dt$ for $j \in J$. Thus, by Lemma 1, we see that $y_* := (x_*, y_{i_*}, \dots, y_{(k+l)_*})$ corresponding u_* minimizes $\Gamma_0(y)$ over all admissible processes (y, u) for (P') with x being sufficiently close to x_* in the norm of L^∞ .

By [4, Theorem 2], we see that there exist Lagrange multipliers $\delta := (\delta_0, \dots, \delta_l) \geq 0$, $x^* \in C^*([0, 1], R^m)$, and $y_i^* \in C^*([0, 1], R)$ $i = 1, \dots, k+l$ not all zero such that

$$(5) \quad 0 \in \partial_y \mathcal{L}(y_*, y^*, u_*, \kappa) + N_{\mathcal{S}}(y_*)$$

$$(6) \quad \mathcal{L}(y_*, y^*, u_*, \kappa) = \min_{u \in \mathcal{U}} \mathcal{L}(y_*, y^*, u, \kappa)$$

$$(7) \quad \delta_j \Gamma_j(y_*) = 0 \quad j \in J$$

where $\mathcal{L}(y, y^*, u, \kappa) := \sum_{i=0}^l \delta_i \Gamma_i(y) + \langle x^*, L_0(y, u) \rangle + \sum_{i=1}^{k+l} \langle y_i^*, L_i(y, u) \rangle$.

According to the formulas of the Clarke gradients (see [3]), we see that

(i) For any $\xi \in \partial \Gamma_0(y_*)$, there are $\bar{\lambda}_i \geq 0, \nu_i \in \partial f_i(x_*(1))$ for $i \in I$ with $\sum_{i \in I} \bar{\lambda}_i = 1$ such that for any $y \in C([0, 1], R^{n+2k})$

$$\langle \xi, y \rangle = \sum_{i \in I} \bar{\lambda}_i y_i(1) + \sum_{i \in I} \bar{\lambda}_i \langle \nu_i, x(1) \rangle.$$

for every $\xi \in \sum_{i=1}^l \delta_i \Gamma_i(y_*)$, there exist $\nu_{k+j} \in \partial g_j(x_*(1))$ for $j \in J$ such that for any $y \in C([0, 1], R^{n+2k})$

$$\langle \xi, y \rangle = \sum_{j \in J} \delta_j y_i(1) + \sum_{j \in J} \delta_j \langle \nu_{k+j}, x(1) \rangle.$$

Analyzing as in [4], we have the following.

(ii) The above multipliers $x^*, y_1^*, \dots, y_{2k}^*$ can be expressed by pairs of the nonnegative Radon measure and Radon-integrable functions (μ_i, ξ_i) , $i = 0, \dots, 2k$. For every $\xi \in \partial_y \left(\langle x^*, L_0(x_*, u_*) \rangle + \sum_{i=1}^{k+l} \langle y_i^*, L_i(x_*, u_*) \rangle \right)$, there is a Lebesgue measurable function $\eta(\cdot)$ with

$$(8) \quad \begin{aligned} \eta(t) \in \quad & \partial_x \left(\left\langle \int_t^1 \xi_0 d\mu_0, \Phi[t] \right\rangle + \sum_{i \in I} \left\langle \int_t^1 \xi_i d\mu_i, F_i(t, x_*(t), u_*(t)) \right\rangle \right. \\ & \left. + \sum_{j \in J} \left\langle \int_t^1 \xi_{k+i} d\mu_{k+i}, G_i(t, x_*(t), u_*(t)) \right\rangle \right) \text{ a.e.,} \end{aligned}$$

such that for any $y \in C([0, 1], R^{n+2k})$,

$$\langle \xi, y \rangle = \int_0^1 \langle x(t) - x(0), \xi_0 \rangle d\mu_0 + \sum_{i=1}^{k+l} \int_0^1 \langle y_i, \xi_i \rangle d\mu_i - \int_0^1 \langle \eta, x \rangle dt.$$

(iii) For each $\xi \in N_S(y_*)$, there is $\alpha \in N_D(x_*(0))$, such that

$$\langle \xi, y \rangle := \langle \alpha, x(0) \rangle \quad \text{for any } y \in C([0, 1], R^{n+k}).$$

Combining (i), (ii) and (iii), from (5) we see that there are $\bar{\lambda}_i, i = 1, \dots, l; \nu_i, i = 1, \dots, k+l; (\mu_i, \xi_i), i = 0, \dots, k+l, \eta$ and α stated above such that

$$\begin{aligned} 0 = & \sum_{i \in I} \delta_0 \bar{\lambda}_i y_i(1) + \sum_{j \in J} \delta_j y_{k+j}(1) + \sum_{i \in I} \delta_0 \bar{\lambda}_i \langle \nu_i, x(1) \rangle + \sum_{j \in J} \delta_j \langle \nu_{k+j}, x(1) \rangle + \\ & \sum_{i=1}^{k+l} \int_0^1 \langle y_i, \xi_i \rangle d\mu_i + \int_0^1 \langle x(t) - x(0), \xi_0 \rangle d\mu_0 - \int_0^1 \langle \eta, x \rangle dt + \langle \alpha, x(0) \rangle \end{aligned}$$

for any $x \in C([0, 1], R^n)$ and $y_i \in C([0, 1], R), i = 1, \dots, k+l$.

Setting $\lambda_i = \delta_0 \bar{\lambda}_i$ for $i \in I, \lambda_{k+j} := \delta_j$ for $j \in J$ and $p(t) := \int_t^1 \xi_0 d\mu_0$, from the above equation, we see that

$$\begin{aligned} \lambda_i y_i(1) + \int_0^1 \left\langle \int_t^1 \xi_i d\mu_i, \dot{y}_i \right\rangle dt &= 0 \quad (\forall y_i \in AC \text{ with } y_i(0) = 0, i \in I \cup J), \\ \langle \alpha, x(0) \rangle + \sum_{i=1}^{k+l} \lambda_i \langle \nu_i, x(1) \rangle + \int_0^1 \left\langle p(t) - \int_t^1 \eta d\tau, \dot{x} \right\rangle dt &= 0 \quad (\forall x \in AC). \end{aligned}$$

These yield that (refer to the proof of [4, Theorem 3])

$$(9) \quad \begin{aligned} \int_t^1 \xi_i d\mu_i &= -\lambda_i, \quad i = 1, \dots, k+l \\ \dot{p}(t) &= -\eta(t) \text{ a.e., } p(0) = \alpha, p(1) = -\sum_{i=1}^{k+l} \lambda_i \nu_i. \end{aligned}$$

Therefore, (9), (8) and (7) imply (1), (2) and (4)

Here, if $\delta = 0$, then $(\lambda_1, \dots, \lambda_{k+l}) = (y_1^*, \dots, y_{k+l}^*) = 0$. From (1) and (2), we can get $p(\cdot) = 0$. Thus, $y^* = 0$ which contradicts that δ and y^* are not all zero. Hence, we have $(\lambda_1, \dots, \lambda_{k+l}) > 0$.

On other hand, By (6) and (9), we see that

$$\int_0^1 H(t, x_*, p, u_*, \lambda) dt = \max_{u \in \mathcal{U}} \int_0^1 H(t, x_*, p, u, \lambda) dt.$$

Discussing as in the proof of [4, Theorem 3], we can obtain (3). \square

According to the results of [8], we see that the above necessary conditions (1)-(4) (Maximum Principle-type) may fail to be sufficient conditions for weak-efficient solutions of (P) even in the "convex" case given below. Next, we give another type necessary weakly-efficiency conditions for (P), which is an extension of [8]. In the "convex" case, the latter necessary conditions are necessary-sufficient for weakly-efficiency under Slater constraint qualifications. Moreover, these conditions are also necessary-sufficient for efficient solutions of (P) under further assumptions.

We impose the following assumption, in which the process (x_*, u_*) will be assumed to be a weakly-efficient solution of type (II) for (P) .

(A5): $F_i(\cdot, x, u)$, $G_i(\cdot, x, u)$, $i = 1, \dots, k$, $\Phi(\cdot, x, u)$ are Lebesgue measurable, and there exist $\epsilon > 0$ and $h_i(t) \in L^1([0, 1], R)$, $i = 0, \dots, k+l$, such that

$$\begin{aligned} |F_i(t, x, u) - F_i(t, x', u')| &\leq h_i(t) (|x - x'| + |u - u'|) \quad \text{for } i \in I \\ |G_j(t, x, u) - G_j(t, x', u')| &\leq h_{k+j}(t) (|x - x'| + |u - u'|) \quad \text{for } j \in J \\ |\Phi(t, x, u(t)) - \Phi(t, x', u'(t))| &\leq h_0(t) (|x - x'| + |u - u'|) \end{aligned}$$

whenever $x, x' \in x_*(t) + \epsilon B_n$, $u, u' \in u_*(t) + \epsilon B_m$ a.e..

Theorem 2: Assume that (A1), (A2) and (A5) be satisfied. Let (x_*, u_*) be a local weakly efficient solution of type (II) for (P) . Then there exist $\lambda = (\lambda_1 \dots, \lambda_{k+l}) > 0$, an absolutely continuous function $p(\cdot) : [0, 1] \rightarrow R^n$ and an integrable function $\zeta(\cdot) : [0, 1] \rightarrow R^m$ such that

$$(10) \quad (-\dot{p}(t), \zeta(t)) \in \partial_{(x,u)} H(t, x_*(t), p(t), u_*(t), \lambda) \quad \text{a.e.}$$

$$(11) \quad p(0) \in N_D(x_*(0)), \quad -p(1) \in \sum_{i \in I} \lambda_i \partial f_i(x_*(1)) + \sum_{j \in J} \lambda_{k+j} \partial g_j(x_*(1))$$

$$(12) \quad \zeta(t) \in N_{U(t)}(u_*(t)) \quad \text{a.e.}$$

$$(13) \quad \lambda_{k+j} \left(\int_0^1 G_j[t] dt + g_j(x_*(1)) \right) = 0 \quad \text{for } j \in J$$

where $H(t, x, p, u, \lambda)$ is defined in Theorem 1.

Proof. It is obvious that the scalar optimization problem in Lemma 1 can be rewritten as follows

$$\begin{aligned} (P^\dagger) : \quad &\text{minimize : } \Gamma(y(1)) := \max_{i \in I, j \in J} \{y_i(1) + f_i(x(1)) - \mathcal{F}_i(x_*, u_*) \\ &\quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad y_{k+j}(1) + g_j(x(1))\} \\ &\text{subject to : } \dot{x}(t) = \Phi(t, x(t), u(t)) \quad \text{a.e.} \\ &\quad \quad \quad \dot{y}_i(t) = F_i(t, x(t), u(t)) \quad \text{a.e.} \quad i \in I \\ &\quad \quad \quad \dot{y}_{k+i}(t) = G_i(t, x(t), u(t)) \quad \text{a.e.} \quad i \in I \\ &\quad \quad \quad x(0) \in C, \quad y_i(0) = 0 \quad i = 1, \dots, 2k, \\ &\quad \quad \quad u(t) \in U(t) \quad \text{a.e.} \end{aligned}$$

where $y := (x, y_1, \dots, y_{2k}) \in AC([0, 1], R^{m+2k})$ is the state and $u \in M([0, 1], R^n)$ is the control.

Define y_* as in proof of Theorem 1. By Lemma 1, we see that (y_*, u_*) is a minimizer over all admissible process for (P^\dagger) with $x(t) \in x_*(t) + \epsilon B_n$, $u(t) \in u_*(t) + \epsilon B_m$ a.e. for some $\epsilon > 0$. Thus, by [8, Proposition 6.1], there exist an absolutely continuous function $\bar{p} = (p, p_1, \dots, p_{k+l})$ and an integrable function ζ such that (12) and the following hold

$$(14) \quad (-\dot{\bar{p}}(t), \dot{y}(t), \zeta(t)) \in \partial_{(y, \bar{p}, u)} \bar{H}(t, y_*(t), \bar{p}(t), u_*(t)) \quad \text{a.e.}$$

$$(15) \quad \bar{p}(0) \in N_{C \times \underbrace{\{0\} \times \dots \times \{0\}}_{2k}}(y_*(0))$$

$$(16) \quad -\bar{p}(1) \in \partial\Gamma(y_*(1))$$

where $\bar{H}(t, y, \bar{p}, u) := \langle p, \Phi(t, x, u) \rangle + \sum_{i \in I} \langle p_i, F_i(t, x, u) \rangle + \sum_{i \in I} \langle p_{k+i}, G_i(t, x, u) \rangle$.

First, let us discuss inclusion (16). Notice that for every $i \in I$ and $j \in J$,

$$\begin{aligned} \Gamma_i(y(1)) &:= y_i(1) + f_i(x(1)) - \mathcal{F}_i(x_*, u_*), \\ \Gamma_j(y(1)) &:= y_{k+j}(1) + g_j(x(1)) \end{aligned}$$

only contains the arguments x and y_i , and $\Gamma_i(y_*(1)) = \Gamma(y_*(1)) = 0$. So by the formulas of the Clarke gradients, there are $\gamma_i \in \partial_x f_i(x_*(1))$ for $i \in I$, $\gamma_{k+j} \in \partial_x g_j(x_*(1))$ for $j \in J$ and $(\lambda_1, \dots, \lambda_{k+l}) > 0$ such that

$$(17) \quad -p(1) = \sum_{i \in I} \lambda_i \gamma_i, \quad -p_i(1) = \lambda_i, \quad i = 1, \dots, k+l.$$

where we can set $\lambda_j = 0$ for $j \in \{j \in J : \mathcal{G}_j(x_*, u_*) < 0\}$.

Thus, (11) and (13) follow from (15) and (17).

On the other hand, since \bar{H} does not contain the arguments y_i , $i = 1, \dots, k+l$, (14) implies that $\dot{p}_i(\cdot) = 0$, $i = 1, \dots, k+l$. Thus, $p_i(\cdot) = -\lambda_i$, $i = 1, \dots, k+l$ and

$$(-\dot{p}(t), \dot{x}(t), \zeta(t)) \in \partial_{(x, \bar{p}, u)} \left(\langle p(t), \Phi[t] \rangle - \sum_{i \in I} \lambda_i F_i[t] - \sum_{i \in I} \lambda_{k+i} G_i[t] \right) \quad a.e.$$

From this inclusion, by the definition of the Clarke generalized gradients, we can easily deduce (10).

Next, we proceed to the optimality conditions for the following problem.

$$\begin{aligned} (P^*) : \quad \min : & \mathcal{F}(x, u) \\ \text{s. t. :} & \dot{x}(t) = A(t)x(t) + B(t)u(t) + b(t) \quad a.e. \\ & x(0) \in D, \quad u(t) \in U(t) \quad a.e. \\ & \mathcal{G}(x, u) \leq 0 \end{aligned}$$

where $x(\cdot) \in AC([0, 1], R^m)$ and $u(\cdot) \in L^1([0, 1], R^n)$, \mathcal{F} and \mathcal{G} are given above, $A(\cdot) : [0, 1] \rightarrow R^{n \times n}$, $B(\cdot) : [0, 1] \rightarrow R^{n \times m}$ are integrable, $b(\cdot) : [0, 1] \rightarrow R^n$ is measurable.

We impose the following hypotheses:

(H1): For every $i \in I$, $F_i(\cdot, x(\cdot), u(\cdot))$ and $G_i(\cdot, x(\cdot), u(\cdot))$ are integrable for any $(x, u) \in AC \times L^1$.

(H2): $F_i(t, \cdot, \cdot)$ for $i \in I$ and $G_j(t, \cdot, \cdot)$ for $j \in J$ are convex lower semicontinuous, and there are $v_i(t) \in L^\infty([0, 1], R^{m+n})$ and $w_i(t) \in L^1([0, 1], R)$, $i = 1, \dots, k+l$ such that for any $x \in R^m$, $u \in R^n$, $F_i(t, x, u) \geq \langle v_i(t), (x, u) \rangle + w_i(t)$ for $i \in I$ and $G_j(t, x, u) \geq \langle v_j(t), (x, u) \rangle + w_j(t)$ for $j \in J$ a.e..

(H3): The functions $f_i(\cdot)$ for $i \in I$ and $g_j(\cdot)$ for $j \in J$ are proper convex and lower semicontinuous.

(H4): The set C is convex, $U(t)$ is convex a.e., and there is $\rho(t) \in L^1$ such that $|u| \leq \rho(t)$ for any $u \in U(t)$ a.e..

(H5): There exists an admissible process (x_i, u_i) for (P^*) , such that $\mathcal{G}_j(x_i, u_i) - \mathcal{G}_j(x_*, u_*) < 0$ for any $j \in \{j \in J : \mathcal{G}_j(x_*, u_*) = 0\}$.

Here, (x_*, u_*) will be assumed to be an admissible process for (P^*) .

Theorem 3: Assume that (H1)-(H5) and (A1) be satisfied. An admissible process (x_*, u_*) is a weakly-efficient solution for (P^*) if and only if there exist $\lambda = (\lambda_1, \dots, \lambda_{k+l}) \geq 0$ with $(\lambda_1, \dots, \lambda_k) > 0$, $p(\cdot) \in AC([0, 1], R^m)$, and $\zeta(\cdot) \in L^\infty([0, 1], R^n)$ such that

$$(18) \quad (\dot{p}(t) + p(t)A(t), p(t)B(t) - \zeta(t)) \in \partial_{(x,u)} \left(\sum_{i \in I} \lambda_i F_i[t] + \sum_{j \in J} \lambda_{k+j} G_j[t] \right) \quad a.e.$$

$$(19) \quad p(0) \in N_C(x_*(1)), \quad -p(1) \in \sum_{i \in I} \lambda_i \partial f_i(x_*(1)) + \sum_{j \in J} \lambda_{k+j} \partial g_j(x_*(1))$$

$$(20) \quad \zeta(t) \in N_{U(t)}(u_*(t)) \quad a.e.,$$

$$(21) \quad \lambda_{k+j} \left(\int_0^1 G_j[t] dt + g_j(x_*(1)) \right) = 0 \quad \text{for } j \in J.$$

Proof. [Necessity] By Lemma 3, we know that there exists $i \in I$ such that (x_*, u_*) is an optimal solution for the following scalar optimal control problem,

$$\begin{aligned} & \text{minimize : } \mathcal{F}_i(x, u) \\ & \text{subject to : } \dot{x}(t) - A(t)x(t) - B(t)u(t) - b(t) = 0 \quad a.e. \\ & \quad \mathcal{G}_j(x, u) \leq 0 \quad j \in J \\ & \quad \mathcal{F}_j(x, u) \leq 0 \quad j \in I/\{i\} \\ & \quad x \in \{x \in AC([0, 1], R^m) : x(0) \in D\} \\ & \quad u \in \mathcal{C} := \{u \in L^1([0, 1], R^n) : u(t) \in U(t) \text{ a.e.}\}. \end{aligned}$$

This means that $(x_*, u_*, x_*(0), x_*(1))$ is a minimizer for the following scalar optimization problem.

$$\begin{aligned} & \text{minimize : } \Lambda_i(z, u, \alpha, \beta) := \int_0^1 F_i(t, z, u) dt + f_i(\beta) \\ & \text{subject to : } \Gamma_1(z, u, \alpha, \beta) := z(t) - \alpha - \int_0^t (Az + Bu + b) d\tau = 0 \quad a.e. \\ & \quad \Gamma_2(z, u, \alpha, \beta) := \beta - \alpha - \int_0^1 (Az + Bu + b) d\tau = 0 \\ & \quad \Lambda_j(z, u, \alpha, \beta) := \int_0^1 F_j(t, z, u) dt + f_j(\beta) - \mathcal{F}_j(x_*, u_*) \leq 0 \text{ for } j \in I/\{i\} \\ & \quad \Lambda_j(z, u, \alpha, \beta) := \int_0^1 G_j(t, z, u) dt + g_j(\beta) \leq 0 \text{ for } j \in J \\ & \quad (z, u, \alpha, \beta) \in \mathcal{M} := L^1([0, 1], R^m) \times \mathcal{C} \times D \times R^m, \end{aligned}$$

where $(z, u, \alpha, \beta) \in L^1([0, 1], R^m) \times L^1([0, 1], R^n) \times R^m \times R^m$

Put $\theta := (z, u, \alpha, \beta)$ and $\theta_* := (x_*, u_*, x_*(0), x_*(1))$. It is obvious that $\Lambda_i(\theta)$ is convex, $\Gamma_1(\theta)$ and $\Gamma_2(\theta)$ are affine mappings. By [5, Theorem 5 p74], there exist $\lambda := (\lambda_1, \dots, \lambda_{k+l}) \geq 0$, $q(\cdot) \in (L^1)^*$ and $\sigma \in R^m$ not all zero, such that

$$(22) \quad \begin{aligned} & \sum_{j=1}^{k+l} \lambda_j \Lambda_j(\theta_*) + \int_0^1 \langle q, \Gamma_1(\theta_*) \rangle dt + \langle \sigma, \Gamma_2(\theta_*) \rangle \\ & = \min_{\theta \in \mathcal{M}} \left(\sum_{j=1}^{k+l} \lambda_j \Lambda_j(\theta) + \int_0^1 \langle q, \Gamma_1(\theta) \rangle dt + \langle \sigma, \Gamma_2(\theta) \rangle \right), \end{aligned}$$

$$\lambda_{k+j}\Lambda_j(\theta_*) = \lambda_{k+j} \left(\int_0^1 G_j[t]dt + g_j(x_*(1)) \right) = 0 \quad \text{for } j \in J$$

Let $I_{\mathcal{M}}(\theta)$ denote the indicator function of \mathcal{M} . Notice that the functions $I_{\mathcal{M}}$, Λ_j ($j \in I$), $\int_0^1 \langle p, \Gamma_1 \rangle dt$, $\langle \sigma, \Gamma_2 \rangle$ are proper convex and lower semicontinuous, from (22) we see that

$$(23) \quad 0 \in \sum_{j=1}^{k+l} \lambda_j \partial \Lambda_j(\theta_*) + \partial \int_0^1 \langle q, \Gamma_1(\theta_*) \rangle dt + \partial \langle \sigma, \Gamma_2(\theta_*) \rangle + N_{\mathcal{M}}(\theta_*)$$

(refer to Section 1 of Chapter 1 in [1]).

Now, we analyze (23). By the formulas of subdifferential (see [1], [5]), we have the following conclusions.

For every $\xi \in \sum_{j=1}^{k+l} \lambda_j \partial \Lambda_j(\theta_*)$, there are $(\mu_j, \eta_j) \in L^\infty$ with $(\mu_j(t), \eta_j(t)) \in \partial_{(x,u)} F_i[t]$ and $\nu_j \in \partial f_j(x_*(1))$ for $j \in I$, $(\mu_{k+j}, \eta_{k+j}) \in L^\infty$ with $(\mu_{k+j}(t), \eta_{k+j}(t)) \in \partial_{(x,u)} G_i[t]$ and $\nu_{k+j} \in \partial g_j(x_*(1))$ for $j \in J$ such that for any $\theta \in L^1 \times L^1 \times R^m \times R^m$

$$\langle \xi, \theta \rangle = \sum_{j=1}^{k+l} \lambda_j \left(\int_0^1 (\langle \mu_j, x \rangle + \langle \eta_j, u \rangle) dt + \langle \nu_j, \beta \rangle \right).$$

Corresponding to any $\xi \in N_{\mathcal{M}}(\theta_*)$, there are $\gamma \in N_D(x_*(0))$, and $\zeta(\cdot) \in N_C(u_*(\cdot))$ such that for any $\theta \in L^1 \times L^1 \times R^m \times R^m$, one has

$$\langle \xi, \theta \rangle = \langle \gamma, \alpha \rangle + \int_0^1 \langle \zeta, u \rangle dt.$$

Notice that $\int_0^1 \langle q, \Gamma_1(\theta) \rangle dt$ is affine on θ , thus $\partial \int_0^1 \langle q, \Gamma_1(\theta_*) \rangle dt = \{\xi\}$ with

$$\langle \xi, \theta \rangle = \int_0^1 \left\langle q, z - \alpha - \int_0^t (Az - Bu) d\tau \right\rangle dt$$

for any $\theta \in L^1 \times L^1 \times R^m \times R^m$.

Similarly, $\partial \langle \sigma, \Gamma_2(\theta_*) \rangle = \{\xi\}$ with

$$\langle \xi, \theta \rangle = \left\langle \sigma, \beta - \alpha - \int_0^1 (Az - Bu) dt \right\rangle$$

for any $\theta \in L^1 \times L^1 \times R^m \times R^m$.

Then, (23) implies that there are (μ_j, η_j) , ν_j , $j = 1, \dots, k+l$, γ and ζ stated above such that

$$(24) \quad \sum_{j=1}^{k+l} \lambda_j \int_0^1 (\langle \mu_j, z \rangle + \langle \eta_j, u \rangle) dt + \sum_{j=1}^{k+l} \lambda_j \langle \nu_j, \beta \rangle + \int_0^1 \left\langle q, z - \int_0^t (Az + Bu) d\tau \right\rangle dt - \left\langle \int_0^1 q dt, \alpha \right\rangle + \left\langle \sigma, \beta - \alpha - \int_0^1 (Az + Bu) dt \right\rangle + \langle \gamma, \alpha \rangle + \int_0^1 \langle \zeta, u \rangle dt = 0$$

for any $(z, u, \alpha, \beta) \in L^1 \times L^1 \times R^m \times R^m$.

Put $p(t) := \int_t^1 q(\tau) d\tau + \sigma$. From (24) we see that

$$\int_0^1 \left\langle \sum_{i=1}^{k+l} \lambda_i \mu_i, z \right\rangle dt - \int_0^1 \langle \dot{p} + pA, z \rangle dt + \int_0^1 \left\langle \sum_{i=1}^{k+l} \lambda_i \eta_i, u \right\rangle dt - \int_0^1 \langle pB - \zeta, u \rangle dt + \left\langle \sum_{i=1}^{k+l} \lambda_i \nu_i, \beta \right\rangle + \langle \sigma, \beta \rangle - \left\langle \int_0^1 q dt, \alpha \right\rangle - \langle \sigma, \alpha \rangle + \langle \gamma, \alpha \rangle = 0$$

for any $(z, u, \alpha, \beta) \in L^1 \times L^1 \times R^m \times R^n$, which implies that

$$(25) \quad \begin{aligned} \dot{p} + pA &= \sum_{i=1}^{k+l} \lambda_i \mu_i, \quad pB - \zeta = \sum_{i=1}^{k+l} \lambda_i \eta_i, \\ p(1) = \sigma &= - \sum_{i=1}^{k+l} \lambda_j \nu_j, \quad p(0) = \int_0^1 q(\tau) d\tau + \sigma = \gamma. \end{aligned}$$

From (25), we obtain (18) and (19).

By $\zeta(\cdot) \in N_C(u_*(\cdot))$, we have $\zeta(t)(u(t) - u_*(t)) \leq 0$ for any $u(\cdot) \in \mathcal{U}$. Thus, from the theory of measurable selection (20) follows.

Finally, if $\lambda = 0$, then (28) and (29) imply that $\sigma = 0$ and $p(\cdot) = 0$, thus λ , q and σ all are zero. Hence, $\lambda > 0$. If $(\lambda_1, \dots, \lambda_k) = 0$, then $(\lambda_k, \dots, \lambda_{k+l}) > 0$. By the Slater constraint qualifications (H5) and the conditions (18)-(21), we have that

$$\begin{aligned} 0 &> \sum_{j \in J} \lambda_{k+j} (\mathcal{G}_j(x_i, u_i) - \mathcal{G}_j(x_*, u_*)) \\ &= \sum_{j \in I/\{i\}} \lambda_j \left(\int_0^1 (G_j(t, x_i, u_i) - G_j[t]) dt + g_j(x_i(1)) - g_j(x_*(1)) \right) \\ &\geq \int_0^1 (\langle \dot{p} + pA, x_i - x_* \rangle + \langle pB - \zeta, u_i - u_* \rangle) dt - p(1)(x_i(1) - x_*(1)) \\ &= -p(0)(x_i(0) - x_*(0)) - \int_0^1 \langle \zeta, u_i - u_* \rangle dt \\ &\geq 0, \end{aligned}$$

a contradiction. Hence, $(\lambda_1, \dots, \lambda_k) > 0$.

[Sufficiency] Assume that there exist $(\lambda_1, \dots, \lambda_k) > 0$, $p(\cdot) \in AC$, and $\zeta(\cdot) \in L^\infty$ satisfying (18)-(21). Notice that $\sum_{i \in I} \lambda_i > 0$, so we can set $\sum_{i \in I} \lambda_i = 1$. Let (x, u) be an arbitrary admissible process for (P^*) . Using (18)-(21) again, we see that

$$\begin{aligned} & \max \{ \mathcal{F}_i(x, u) - \mathcal{F}_i(x_*, u_*) : i \in I \} \\ & \geq \sum_{i \in I} \lambda_i \left(\int_0^1 F_i(t, x, u) dt + f_i(x(1)) - \int_0^1 F_i[t] dt - f_i(x_*(1)) \right) \\ & \quad \sum_{j \in J} \lambda_{k+j} \left(\int_0^1 G_j(t, x, u) dt + g_j(x(1)) - \int_0^1 G_j[t] dt - g_j(x_*(1)) \right) \\ & \quad + \int_0^1 \langle p, \dot{x} - Ax - Bu - b \rangle dt - \int_0^1 \langle p, \dot{x}_* - Ax_* - Bu_* - b \rangle dt \\ & = \int_0^1 \left(\sum_{i \in I} \lambda_i F_i(t, x, u) + \sum_{j \in J} \lambda_{k+j} G_j(t, x, u) dt - \sum_{i \in I} \lambda_i F_i[t] - \sum_{j \in J} \lambda_{k+j} G_j[t] \right) dt \\ & \quad + \sum_{i \in I} \lambda_i f_i(x(1)) + \sum_{j \in J} \lambda_{k+j} g_j(x(1)) - \sum_{i \in I} \lambda_i f_i(x_*(1)) - \sum_{j \in J} \lambda_{k+j} g_j(x_*(1)) \\ & \quad - \int_0^1 (\langle \dot{p} + pA, x - x_* \rangle + \langle pB - \zeta, u - u_* \rangle) dt - \int_0^1 \langle \zeta, u - u_* \rangle dt \\ & \quad + \langle p(1), x(1) - x_*(1) \rangle - \langle p(0), x(0) - x_*(0) \rangle \\ & \geq 0. \end{aligned}$$

By Lemma 1, (x_*, u_*) is a weakly-efficient solution for (P) . \square

Using Theorem 3 and Lemma 3, we can easily show that the conditions (18)-(21) in Theorem 3 are also necessary-sufficient for efficient solutions of (P^*) under the following Slater constraint qualifications (H6).

(H6): For every $i \in I$, there is an admissible process (x_i, u_i) for (P^*) , such that $\mathcal{F}_j(x_i, u_i) - \mathcal{F}_i(x_*, u_*) < 0$ for any $j \in I/\{i\}$ and $\mathcal{G}_j(x_i, u_i) - \mathcal{G}_j(x_*, u_*) < 0$ for any $j \in \{j \in J : \mathcal{G}_j(x_*, u_*) = 0\}$

Theorem 4: Assume that (H1)-(H6) and (A1) are satisfied. An admissible process (x_*, u_*) is an efficient solution for (P^*) if and only if there exist $(\lambda_1 \dots, \lambda_{k+1}) \geq 0$ with $(\lambda_1 \dots, \lambda_k) \gg 0$, $p(\cdot) \in AC([0, 1], R^m)$, and $\zeta(\cdot) \in L^\infty([0, 1], R^n)$ such that (18)-(21) hold.

Remark. It is easy to see that the sufficiency in Theorem 3 and Theorem 4 also hold under the following simpler assumptions: F_i for $i \in I$ and G_j for $j \in I$ are convex in (x, u) and measurable in t , f_i for $i \in I$ and g_j for $j \in I$ are convex functions, C is convex set and $U(t)$ is convex a.e..

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