

Necessary Conditions for Optimization in Multiparameter Discrete Systems¹

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Abstract. A general first-order dynamic representation for discrete systems with several independent variables is proposed, based on the Dieudonné–Rashevsky form for partial differential equations. This representation does not restrict consideration to causal systems. A minimum principle for such systems is proved, thus extending results known for discrete-time systems to the case of several independent variables. The proof requires only the classical implicit function theorem.

Key Words. Discrete models, partial difference equations, Dieudonné–Rashevsky form, noncausal systems, necessary conditions, derived cones, separation of convex sets, implicit function theorem.

1. Introduction

The widespread availability of high-speed digital computers has sparked a renewal of interest in the use of difference equations for modeling dynamic systems. At the same time, researchers in diverse scientific disciplines are realizing that, in many cases, difference equations give a much more accurate representation of system dynamics than differential equations. Numerous authoritative treatises defending the use of discrete mathematical models have appeared (Refs. 1–3), as have sophisticated

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analyses of the dynamics of discrete models (Refs. 4–6). Moreover, advances in electronics are creating new dynamic systems for which a continuum mathematical model is decidedly inappropriate (Ref. 7). Until very recently, research in discrete system theory focused on systems with only one independent variable. However, the capability for high-speed digital signal processing (in particular, of two-dimensional images) has generated much interest in the development both of discrete models in several (in particular, two) independent variables (Refs. 8–10) and of associated discrete multiparameter system theory (Ref. 11).

In this paper, a general model for discrete multiparameter systems consisting of first-order nonlinear partial difference equations is proposed. This model is a discrete form of the Dieudonné–Rashevsky (DR) equations used by Cesari (Ref. 12) and Suryanarayana (Ref. 13) in optimization studies with partial differential equations. The model has descriptor form (see Ref. 14), making explicit a property used by Cesari and Suryanarayana without comment. Most systems of partial difference equations encountered in applications (including noncausal systems) can be placed in this discrete DR form (Ref. 15). In particular, most state-space system models recently developed for linear two-parameter discrete (2-D) systems (Refs. 16–18) can be placed in this form.

The main contribution of this paper is the statement and proof of necessary conditions for optimal control of multiparameter systems in the discrete DR form. The approach to the proof is motivated by the work of Baum and Cesari (Ref. 19) in simplifying the proof of the Pontryagin maximum principle. A notable feature is that only the classical implicit function theorem (rather than a fixed-point theorem) is required. The theory of optimization for discrete-time systems is mature (Refs. 20–25). Optimization with multiparameter systems involving discrete variables has proceeded in several directions with respect to the type of models used: discretization in time only (e.g., Ref. 26); discretization in space variables only (e.g., Ref. 27); and discretization in all variables (Refs. 28–31). Results in the present paper extend to fully discrete multiparameter systems the sharpest general minimum principle known for discrete-time systems (Ref. 25).

2. Optimization Problem

Problem Definition. The optimization problem is in Mayer form with partial difference equations and static constraints defined over a finite rectangular array of multiindices. Given positive integers m and k_α , $\alpha = 1, \dots, m$, denote by G the Cartesian product (array) $\prod_{\alpha=1}^m \{0, \dots, k_\alpha\}$, and

by k the corner point (k_1, \dots, k_m) . For each index α , denote by 1^α the unit displacement $(0, \dots, 0, 1, 0, \dots, 0)$ in the α th variable, and by $G_k(\alpha)$ the subset

$$\{i \equiv (i_1, \dots, i_m) \in G : i_\alpha \neq k_\alpha\}.$$

Assume that the following are specified: additional positive integers s_α , $\alpha = 1, \dots, m$, and n ; a nonempty set U_0 ; a family of subsets $U(i)$, $i \in G \sim k$, of U_0 (the symbol \sim denotes set-theoretic difference); a family of subsets $X(i)$, $i \in G$, of Euclidean space E^n ; a family of real matrices $S^\alpha(i)$: $s_\alpha \times n$, $i \in G_k(\alpha)$, $\alpha = 1, \dots, m$; a family of functions $f^\alpha(i, \cdot, \cdot)$: $E^n \times U_0 \rightarrow E^{s_\alpha}$, $i \in G_k(\alpha)$, $\alpha = 1, \dots, m$; and a function f_0 : $E^n \rightarrow E^1$. A pair (x, u) , with $x: G \rightarrow E^n$ and $u: G \sim k \rightarrow U_0$ is called *admissible* if it satisfies:

$$\begin{aligned} S^\alpha(i)x(i+1^\alpha) &= f^\alpha(i, x(i), u(i)), & i \in G_k(\alpha), \alpha = 1, \dots, m, \\ u(i) &\in U(i), & i \in G \sim k, \\ x(i) &\in X(i), & i \in G. \end{aligned} \tag{1}$$

The optimization problem is to minimize the functional

$$(x, u) \mapsto I(x, u) \triangleq f_0(x(k))$$

over the class of admissible pairs. A minimizing pair is called *optimal*.

Discrete System Model. Equations (1) constitute the discrete Dieudonné-Rashevsky form for multiparameter partial difference equations. Its relationship with proposed multiparameter state-space models for linear systems is briefly indicated for the 2-D case. Kung *et al.* (Ref. 32) have shown that the Roesser model (Ref. 17) is the most general of the models proposed in Refs. 16-18 (see Ref. 33 for an alternative viewpoint). Consider the 2-D Roesser model with horizontal and vertical local states $x_1: n_1 \times 1$ and $x_2: n_2 \times 1$, respectively (using the notation of the preceding paragraph),

$$\begin{aligned} x_1(i+(1, 0)) &= A_1 x_1(i) + B_1 x_2(i) + C_1 u(i), & i \equiv (i_1, i_2) \in G_k(1), \\ x_2(i+(0, 1)) &= A_2 x_1(i) + B_2 x_2(i) + C_2 u(i), & i \in G_k(2). \end{aligned}$$

See Ref. 17 for the distinction between local and global states in 2-D systems. These equations are easily expressed in the form (1) for the vector

$$x \triangleq (x_1, x_2)^T$$

by using

$$S^1 \triangleq \begin{bmatrix} I_{n_1} & 0 \\ 0 & 0 \end{bmatrix} \quad \text{and} \quad S^2 \triangleq \begin{bmatrix} 0 & 0 \\ 0 & I_{n_2} \end{bmatrix}.$$

Here, I_ρ denotes the identity matrix, and the superscript T denotes matrix transpose. Recently, Chan (Ref. 34) has introduced a new multiparameter state-space model in which more general parameters replace the unit shifts 1^α . This model can be translated into a form similar to Eqs. (1) if Chan's generalized shifts are used in (1) instead of the shifts 1^α . More importantly, it should be observed that the proposed form (1) is not restricted to use with causal systems as are all of the local state-space models. Some of the fundamental difficulties associated with the development of multiparameter discrete linear system theory are outlined in Ref. 35. However, a discussion of these matters, as they relate to the linear form of Eqs. (1), is beyond the scope of the present paper.

3. Necessary Conditions

Preliminaries. Several definitions facilitate the statement of results. With each point

$$i \equiv (i_1, \dots, i_m)$$

of the array G are associated the index sets

$$\Gamma^j(i) \triangleq \{\alpha \in \{1, \dots, m\} : i_\alpha \neq j_\alpha\}, \quad j = 0, k;$$

and with each index α in $\{1, \dots, m\}$ is associated the subset

$$G_0(\alpha) \triangleq \{i \equiv (i_1, \dots, i_m) \in G : i_\alpha \neq 0\}.$$

The boundary of G is the set

$$B \triangleq \{i \equiv (i_1, \dots, i_m) \in G : \exists \alpha \in \{1, \dots, m\} \text{ with } i_\alpha \in \{0, k_\alpha\}\}.$$

A *Hamiltonian function*, defined for i in $G \sim k$, x in E^n , u in $U(i)$, p^α in E^{s_α} , $\alpha = 1, \dots, m$, is given by

$$H(i, x, u, p^1, \dots, p^m) \triangleq \sum_{\alpha \in \Gamma^k(i)} p^\alpha \cdot f^\alpha(i, x, u).$$

Here, the symbol \cdot denotes scalar product.

Given a nonempty set A in E^n and a point x of A , a closed convex cone C is said to be a *derived cone* for A at x if, for each finite subset h_ρ , $\rho = 1, \dots, r$, of C , there exists a continuously differentiable map $\sigma : [0, 1]^r \rightarrow E^n$ such that:

- (i) $\sigma(t) \in A$, for each $t \equiv (t_1, \dots, t_r)$ in $[0, 1]^r$;
- (ii) $\sigma(0) = x$;
- (iii) $(\partial\sigma/\partial t_\rho)(0) = h_\rho$, $\rho = 1, \dots, r$.

The polar of a closed convex cone in E^n is denoted by the superscript $^\circ$.

Theorem 3.1. Minimum Principle. Assume the following:

(H1) f_0 is continuously differentiable on (i.e., in a neighborhood of) $X(k)$;

(H2) for each triple α in $\{1, \dots, m\}$, i in $G_k(\alpha)$, u in $U(i)$, the map $f^\alpha(i, \cdot, u)$ is continuously differentiable on $X(i)$;

(H3) for each pair i in $G \sim k$, x in $X(i)$, the set $f(i, x, U(i))$, where f denotes $(f^\alpha)_{\alpha \in \Gamma^k(i)}$, is convex.

Let (x^*, u^*) be an optimal pair; and, for each i in G , let $C(i)$ be a derived cone for $X(i)$ at $x^*(i)$.

Then, there exist $p_0 \geq 0$ and vectors $p^\alpha: G_k(\alpha) \rightarrow E^{s_\alpha}$, $\alpha = 1, \dots, m$, satisfying:

- (a) $(p_0, p^1, \dots, p^m) \neq 0$;
- (b) $\sum_{\alpha=1}^m S^{\alpha T} (i - 1^\alpha) p^\alpha (i - 1^\alpha) - (\partial H / \partial x)^T (i, x^*(i), u^*(i), p^1(i), \dots, p^m(i)) \in [C(i)]^\circ, i \in G \sim B$;
- (c) $H(i, x^*(i), u^*(i), p^1(i), \dots, p^m(i)) = \min_{u \in U(i)} \{H(i, x^*(i), u, p^1(i), \dots, p^m(i))\}, i \in G \sim k$;
- (d) $\sum_{\alpha \in \Gamma^0(i)} S^{\alpha T} (i - 1^\alpha) p^\alpha (i - 1^\alpha) - (\partial H / \partial x)^T (i, x^*(i), u^*(i), p^1(i), \dots, p^m(i)) \in [C(i)]^\circ, i \in B \sim k$;
 $\sum_{\alpha=1}^m S^{\alpha T} (k - 1^\alpha) p^\alpha (k - 1^\alpha) - p_0 (\partial f_0 / \partial x)^T (x^*(k)) \in [C(k)]^\circ$.

The sum over the empty set that occurs in (d) is assigned the value zero.

4. Proof of Necessary Conditions

Motivation. Note that Eqs. (1) are equivalent to γ scalar equations, where

$$\gamma \triangleq \sum_{\alpha=1}^m s_\alpha \text{card } G_k(\alpha).$$

The notation $(z_{i\alpha})_{i\alpha}$ is used to denote vectors $(z_{i\alpha})_{i \in G_k(\alpha)}$, $\alpha = 1, \dots, m$, in E^γ . Consider the set \mathcal{P} of elements $(z_0, (z_{i\alpha})_{i\alpha})$ in the augmented space

$E^{\gamma+1}$ defined by the conditions

$$\begin{aligned} z_0 &= f_0(x(k)) - f_0(x^*(k)), \\ z_{i\alpha} &= f^\alpha(i, x(i), u(i)) - S^\alpha(i)x(i+1^\alpha), \quad i \in G_k(\alpha), \alpha = 1, \dots, m, \\ u(i) &\in U(i), \quad i \in G \sim k, \\ x(i) &\in X(i), \quad i \in G, \end{aligned}$$

where (x^*, u^*) represents an admissible pair, and also the set

$$\mathcal{Q} \triangleq \{(-\eta, 0, \dots, 0) \in E^{\gamma+1}; \eta > 0\}.$$

Disjointness of \mathcal{P} and \mathcal{Q} characterizes optimality of a pair (x^*, u^*) . Proof of the necessary conditions follows from establishing the weaker condition that \mathcal{Q} is algebraically separated from a suitable convex approximation to \mathcal{P} .

Proof. For each triple α in $\{1, \dots, m\}$, i in $G_k(\alpha)$, u in $U(i)$, denote

$$\begin{aligned} A^\alpha(i) &\triangleq (\partial f^\alpha / \partial x)(i, x^*(i), u^*(i)), \\ F^\alpha(i, u) &\triangleq f^\alpha(i, x^*(i), u) - f^\alpha(i, x^*(i), u^*(i)), \end{aligned}$$

and also denote

$$a_0 \triangleq (\partial f_0 / \partial x)^T(x^*(k)).$$

Denote by \mathcal{R} the set of points $(z_0, (z_{i\alpha})_{i\alpha})$ in $E^{\gamma+1}$ defined by the conditions

$$z_0 = a_0 \cdot h(k), \tag{2}$$

$$z_{i\alpha} = A^\alpha(i)h(i) + F^\alpha(i, u(i)) - S^\alpha(i)h(i+1^\alpha), \quad i \in G_k(\alpha), \alpha = 1, \dots, m, \tag{3}$$

$$u(i) \in U(i), \quad i \in G \sim k, \tag{4}$$

$$h(i) \in C(i), \quad i \in G. \tag{5}$$

In view of the hypotheses, it follows immediately that \mathcal{R} is convex and contains the origin. To show that \mathcal{Q} and \mathcal{R} are separated, two cases are distinguished, depending on whether or not \mathcal{R} has (topological) interior points. The interior of \mathcal{R} , denoted by $\text{int}(\mathcal{R})$, is convex since \mathcal{R} is convex.

Separation of \mathcal{Q} and \mathcal{R} . If \mathcal{R} has empty interior, then the linear manifold in $E^{\gamma+1}$ generated by \mathcal{R} is a proper affine space that can be extended to a hyperplane, say $p \cdot z = \beta$, for some nonzero p in $E^{\gamma+1}$; the first component, say p_0 , of p can be assumed nonnegative. Since \mathcal{R} contains

the origin and lies in the hyperplane, it follows that $\beta = 0$. Then, for each $z \equiv (-\eta, 0, \dots, 0)$ in \mathcal{Q} ,

$$p \cdot z = -p_0\eta \leq 0,$$

which establishes separation for this trivial case.

If \mathcal{R} has nonempty interior, it can be shown that \mathcal{Q} and the interior of \mathcal{R} are disjoint. Verification of this assertion entails technical arguments peripheral to the main thrust of the proof, and is therefore postponed to later in this section. Assuming that this assertion is true, it follows from well-known separation results that \mathcal{Q} and $\text{int}(\mathcal{R})$ are separated by a hyperplane, say $p \cdot z = \beta$ ($p \neq 0$); in particular,

$$\begin{aligned} p \cdot z &\geq \beta, & \text{for } z \text{ in } \text{int}(\mathcal{R}), \\ p \cdot z &\leq \beta, & \text{for } z \text{ in } \mathcal{Q}. \end{aligned}$$

But, since $\text{int}(\mathcal{R})$ is nonempty,

$$\text{cl}(\mathcal{R}) = \text{cl}(\text{int}(\mathcal{R})),$$

where $\text{cl}(\cdot)$ denotes the closure (Ref. 36). Hence, \mathcal{R} itself lies in the closed half-space

$$p \cdot z \geq \beta.$$

Denote by p_0 the first component of p , and let $\eta > 0$ be arbitrary. Since the zero vector is in \mathcal{R} , and since $(-\eta, 0, \dots, 0)$ is in \mathcal{Q} , the inequalities

$$p_0(-\eta) \leq \beta \leq 0$$

hold. Hence,

$$p_0 \geq 0.$$

Moreover, since η is arbitrary, β must be zero.

Deduction of the Necessary Conditions. The preceding subsection shows that \mathcal{R} lies in the half-space

$$p \cdot z \geq 0,$$

defined by a nonzero vector $p \equiv (p_0, (p^\alpha(i))_{i\alpha})$ satisfying

$$p_0 \geq 0.$$

This proves assertion (a).

Let $u: G \sim k \rightarrow U_0$ and $h: G \rightarrow E^n$ satisfy Eqs. (4) and (5), respectively, but be otherwise arbitrary. This defines a general point of \mathcal{R} given by Eqs.

(2) and (3). The separation inequality for such a point of \mathcal{R} is the following:

$$\begin{aligned}
 & p_0 a_0 \cdot h(k) + \sum_{\alpha=1}^m \sum_{i \in G_k(\alpha)} A^{\alpha T}(i) p^\alpha(i) \cdot h(i) \\
 & + \sum_{\alpha=1}^m \sum_{i \in G_k(\alpha)} p^\alpha(i) \cdot F^\alpha(i, u(i)) \\
 & - \sum_{\alpha=1}^m \sum_{i \in G_k(\alpha)} S^{\alpha T}(i) p^\alpha(i) \cdot h(i + 1^\alpha) \geq 0.
 \end{aligned}$$

Following some algebra, this inequality may be rewritten in a form that isolates the free parameter vectors $u(\cdot)$ and $h(\cdot)$:

$$\begin{aligned}
 & \sum_{i \in G \sim B} \left\{ \sum_{\alpha=1}^m [S^{\alpha}(i - 1^\alpha) p^\alpha(i - 1^\alpha) - A^{\alpha T}(i) p^\alpha(i)] \right\} \cdot h(i) \\
 & + \sum_{i \in B \sim (0, k)} \left[\sum_{\alpha \in \Gamma^0(i)} S^{\alpha T}(i - 1^\alpha) p^\alpha(i - 1^\alpha) \right. \\
 & \left. - \sum_{\alpha \in \Gamma^k(i)} A^{\alpha T}(i) p^\alpha(i) \right] \cdot h(i) \\
 & - \sum_{\alpha \in \Gamma^k(0)} [A^{\alpha T}(0) p^\alpha(0)] \cdot h(0) \\
 & + \left[\sum_{\alpha=1}^m S^{\alpha T}(k - 1^\alpha) p^\alpha(k - 1^\alpha) - p_0 a_0 \right] \cdot h(k) \\
 & \leq \sum_{i \in G \sim k} \sum_{\alpha \in \Gamma^k(i)} p^\alpha(i) \cdot F^\alpha(i, u(i)). \tag{6}
 \end{aligned}$$

Assertions (b)–(d) are deduced by writing Ineq. (6) with several judicious choices of the free parameter vectors.

First, choose

$$h(i) \triangleq 0, \quad i \in G,$$

select arbitrary points i in $G \sim k$ and w in $U(i)$, and choose for u the mapping

$$u(\lambda) \triangleq \begin{cases} u^*(\lambda), & \lambda \in G \sim \{k, i\}, \\ w, & \lambda = i. \end{cases}$$

With this choice of free parameter vectors, Ineq. (6) reduces to

$$\sum_{\alpha \in \Gamma^k(i)} p^\alpha(i) \cdot F^\alpha(i, w) \geq 0,$$

which is equivalent to assertion (c).

Next, choose

$$u(i) \triangleq u^*(i), \quad i \in G \sim k,$$

select arbitrary points i in G and ζ in $C(i)$, and choose for h the mapping

$$h(\lambda) \triangleq \begin{cases} 0, & \lambda \in G \sim i, \\ \zeta, & \lambda = i. \end{cases}$$

With this choice of free parameter vectors, Ineq. (6) reduces to either

$$\left\{ \sum_{\alpha=1}^m [S^{\alpha T} (i-1^\alpha) p^\alpha (i-1^\alpha) - A^{\alpha T} (i) p^\alpha (i)] \right\} \cdot \zeta \leq 0, \quad (7)$$

$$\left[\sum_{\alpha \in \Gamma^0(i)} S^{\alpha T} (i-1^\alpha) p^\alpha (i-1^\alpha) - \sum_{\alpha \in \Gamma^k(i)} A^{\alpha T} (i) p^\alpha (i) \right] \cdot \zeta \leq 0, \quad (8)$$

$$- \left[\sum_{\alpha \in \Gamma^k(0)} A^{\alpha T} (0) p^\alpha (0) \right] \cdot \zeta \leq 0, \quad (9)$$

or

$$\left[\sum_{\alpha=1}^m S^{\alpha T} (i-1^\alpha) p^\alpha (i-1^\alpha) - p_0 a_0 \right] \cdot \zeta \leq 0, \quad (10)$$

according as the point i selected lies in

- (i) $G \sim B$,
- (ii) $B \sim \{0, k\}$,
- (iii) $\{0\}$,
- (iv) $\{k\}$,

respectively. Since ζ is an arbitrary element of $C(i)$, Ineqs. (7)–(10) imply, respectively, that

$$\begin{aligned} & \sum_{\alpha=1}^m [S^{\alpha T} (i-1^\alpha) p^\alpha (i-1^\alpha) - A^{\alpha T} (i) p^\alpha (i)] \in [C(i)]^\circ, \\ & \sum_{\alpha \in \Gamma^0(i)} S^{\alpha T} (i-1^\alpha) p^\alpha (i-1^\alpha) - \sum_{\alpha \in \Gamma^k(i)} A^{\alpha T} (i) p^\alpha (i) \in [C(i)]^\circ, \\ & - \sum_{\alpha \in \Gamma^k(0)} A^{\alpha T} (0) p^\alpha (0) \in [C(i)]^\circ, \end{aligned}$$

or

$$\sum_{\alpha=1}^m S^{\alpha T} (i-1^\alpha) p^\alpha (i-1^\alpha) - p_0 a_0 \in [C(i)]^\circ,$$

which are equivalent, collectively, to assertions (b) and (d). The proof is complete, except for the verification that \mathcal{Q} and $\text{int}(\mathcal{R})$ are disjoint.

Disjointness of \mathcal{Q} and $\text{int}(\mathcal{R})$. The proof is by contraposition. Suppose that

$$z \equiv (z_0, (z_{i\alpha})_{i\alpha}) \in \mathcal{Q} \cap \text{int}(\mathcal{R}).$$

It will be shown that this implies that (x^*, u^*) cannot be optimal. On the one hand, since $z \in \mathcal{Q}$, it follows that

$$\eta \triangleq -z_0 > 0$$

and

$$z_{i\alpha} = 0, \quad \text{for each } i \text{ in } G_k(\alpha), \alpha = 1, \dots, m.$$

On the other hand, since $z \in \text{int}(\mathcal{R})$, there is an open set N in $E^{\gamma+1}$ with $z \in N \subset \mathcal{R}$. Consider the vectors

$$e^\rho \triangleq (0, \dots, 0, 1, 0, \dots, 0), \quad \rho = 1, \dots, \gamma,$$

in $E^{\gamma+1}$, with the unit in the $(\rho + 1)$ th coordinate for each, and the sum vector

$$b \triangleq \sum_{\rho=1}^{\gamma} e^\rho.$$

Then, there exists $\delta > 0$ such that the simplex with vertices $z + \delta e^\rho$, $\rho = 1, \dots, \gamma$, and $z - \delta b$ lies inside N , and hence in \mathcal{R} . Using Eqs. (2)–(5), it follows that:

(i) for each $\rho = 1, \dots, \gamma$, there exist $u^\rho: G \sim k \rightarrow U_0$ and $h^\rho: G \rightarrow E^n$ such that

$$\begin{aligned} z_0 &= a_0 \cdot h^\rho(k), \\ (z_{i\alpha})_{i\alpha} + (0, \dots, 0, \delta, 0, \dots, 0) &= (A^\alpha(i)h^\rho(i) + F^\alpha(i, u^\rho(i)) - S^\alpha(i)h^\rho(i + 1^\alpha))_{i\alpha}, \\ u^\rho(i) &\in U(i), \quad i \in G \sim k, \\ h^\rho(i) &\in C(i), \quad i \in G, \end{aligned}$$

with each δ appearing, respectively, in the ρ th coordinate;

(ii) there exist $u^{\gamma+1}: G \sim k \rightarrow U_0$ and $h^{\gamma+1}: G \rightarrow E^n$ such that

$$\begin{aligned} z_0 &= a_0 \cdot h^{\gamma+1}(k), \\ (z_{i\alpha})_{i\alpha} + (-\delta, \dots, -\delta) &= (A^\alpha(i)h^{\gamma+1}(i) + F^\alpha(i, u^{\gamma+1}(i)) - S^\alpha(i)h^{\gamma+1}(i + 1^\alpha))_{i\alpha}, \\ u^{\gamma+1}(i) &\in U(i), \quad i \in G \sim k, \\ h^{\gamma+1}(i) &\in C(i), \quad i \in G. \end{aligned}$$

Comparison of the two expressions for the components of z shows that

$$a_0 \cdot h^\rho(k) = -\eta, \quad \rho = 1, \dots, \gamma + 1, \tag{11}$$

$$\begin{aligned} & (A^\alpha(i)h^\rho(i) + F^\alpha(i, u^\rho(i)) - S^\alpha(i)h^\rho(i + 1^\alpha))_{i\alpha} \\ &= \begin{cases} (0, \dots, 0, \delta, 0, \dots, 0), & \rho = 1, \dots, \gamma, \\ (-\delta, \dots, -\delta), & \rho = \gamma + 1. \end{cases} \end{aligned} \tag{12}$$

Let $i \in G$ be arbitrary. Since $h^\rho(i) \in C(i)$, $\rho = 1, \dots, \gamma + 1$, and since $C(i)$ is a derived cone for $X(i)$ at $x^*(i)$, there exists a continuously differentiable C^1 -map $\sigma(i; \cdot): [0, 1]^{\gamma+1} \rightarrow E^n$ such that

$$\sigma(i; t) \in X(i), \quad \text{for each } t \equiv (t_1, \dots, t_{\gamma+1}) \text{ in } [0, 1]^{\gamma+1},$$

$$\sigma(i; 0) = x^*(i),$$

and

$$(\partial\sigma/\partial t_\rho)(i, 0) = h^\rho(i), \quad \rho = 1, \dots, \gamma + 1.$$

The map $\sigma(i; \cdot)$ can be extended to a C^1 -map $\hat{\sigma}(i; \cdot)$ on some neighborhood of $[0, 1]^{\gamma+1}$. Let D_1 be a neighborhood of $[0, 1]^{\gamma+1}$ in $E^{\gamma+1}$ on which each extension $\hat{\sigma}(i; \cdot)$, $i \in G$, is C^1 . For each triple $t \equiv (t_1, \dots, t_{\gamma+1})$ in D_1 , α in $\{1, \dots, m\}$, i in $G_k(\alpha)$, define $y^\alpha(i + 1^\alpha; t)$ as the sum

$$\left[1 - \sum_{\rho=1}^{\gamma+1} t_\rho \right] f^\alpha(i, \hat{\sigma}(i; t), u^*(i)) + \sum_{\rho=1}^{\gamma+1} t_\rho f^\alpha(i, \hat{\sigma}(i; t), u^\rho(i)).$$

For $t = 0$, it follows from Eqs. (1) that

$$y^\alpha(i + 1^\alpha; 0) = S^\alpha(i)x^*(i + 1^\alpha), \quad i \in G_k(\alpha), \alpha = 1, \dots, m.$$

For t satisfying

$$t_\rho \geq 0, \quad \rho = 1, \dots, \gamma + 1, \quad \sum_{\rho=1}^{\gamma+1} t_\rho \leq 1, \tag{13}$$

then

$$\hat{\sigma}(i; t) = \sigma(i; t) \in X(i), \quad \text{for each } i \text{ in } G.$$

Hypothesis (H3) of Theorem 3.1 ensures that each of the sets $f(i, \sigma(i; t), U(i))$, $i \in G \sim k$, is convex. Hence, there exist maps $u(\cdot; t): G \sim k \rightarrow U_0$, with $u(i; t) \in U(i)$, for each i in $G \sim k$, such that

$$y^\alpha(i + 1^\alpha; t) = f^\alpha(i, \sigma(i; t), u(i; t)), \quad i \in G_k(\alpha), \alpha = 1, \dots, m.$$

Each map $y^\alpha(i + 1^\alpha; \cdot)$, $i \in G_k(\alpha)$, $\alpha = 1, \dots, m$, is C^1 on some neighborhood of 0 contained in D_1 . On such a neighborhood,

$$(\partial y^\alpha/\partial t_\rho)(i + 1^\alpha; 0) = A^\alpha(i)h^\rho(i) + F^\alpha(i, u^\rho(i)), \quad \rho = 1, \dots, \gamma + 1.$$

Denote by D a neighborhood of 0 in $E^{\gamma+1}$ on which each of the maps $y^\alpha(i+1^\alpha; \cdot), i \in G_k(\alpha), \alpha = 1, \dots, m$, and $f_0(\hat{\sigma}(k; \cdot))$ is C^1 . Define the map

$$\Theta \equiv (\Theta_0, (\Theta_{i\alpha})_{i\alpha}): D \times E^{\gamma+1} \rightarrow E^n$$

by

$$\Theta_0(t_0, t) \triangleq f_0(\hat{\sigma}(k; t)) + t_0 - f_0(x^*(k)),$$

$$\Theta_{i\alpha}(t_0, t) \triangleq y^\alpha(i+1^\alpha; t) - S^\alpha(i)\hat{\sigma}(i+1^\alpha; t), \quad i \in G_k(\alpha), \alpha = 1, \dots, m.$$

Then, Θ is C^1 and

$$\Theta(0, 0) = 0.$$

Moreover, using Eqs. (11)–(12), it is easy to show that

$$(\partial\Theta/\partial t)(0, 0) = (-1)^{\gamma+2}\eta(\gamma+2)\delta^{\gamma+1} \neq 0.$$

By the classical implicit function theorem, there exist $\delta_0 > 0$ and a C^1 -map $T \equiv (T_1, \dots, T_{\gamma+1}): D \rightarrow E^{\gamma+1}$ satisfying

$$T(0) = 0 \quad \text{and} \quad \Theta(t_0, T(t_0)) = 0, \quad -\delta_0 < t_0 < \delta_0. \quad (14)$$

Since T is continuous, there exists $\delta_0^1 > 0, \delta_0^1 \leq \delta_0$, such that, for each ρ in $\{1, \dots, \gamma+1\}$, if

$$0 < t_0 < \delta_0^1,$$

then

$$|T_\rho(t_0)| < (\gamma+1)^{-1}.$$

Differentiating the second of Eqs. (14) and applying Cramer's rule yields expressions for the derivatives of the components of T on $(-\delta_0, \delta_0)$. In particular,

$$(dT_\rho/dt_0)(0) = \eta^{-1}(\gamma+2)^{-1} > 0, \quad \rho = 1, \dots, \gamma+1.$$

It follows from the fundamental theorem of calculus that there exists $\delta_0^2 > 0, \delta_0^2 \leq \delta_0^1$, such that, if

$$0 < t_0 < \delta_0^2,$$

then

$$T_\rho(t_0) > 0, \quad \rho = 1, \dots, \gamma+1.$$

Choose t_0 satisfying

$$0 < t_0 < \delta_0^2.$$

It is clear from the preceding that the components of $T(t_0)$ satisfy relations (13). Hence, there exists a map $u(\cdot; T(t_0)): G \sim k \rightarrow U_0$, with $u(i; T(t_0)) \in U(i)$, $i \in G \sim k$, such that

$$y^\alpha(i + 1^\alpha; T(t_0)) = f^\alpha(i, \sigma(i; T(t_0)), u(i; T(t_0))),$$

$$i \in G_k(\alpha), \quad \alpha = 1, \dots, m.$$

Thus, for each pair α in $\{1, \dots, m\}$, i in $G_k(\alpha)$, the condition

$$\Theta_{i\alpha}(t_0, T(t_0)) = 0$$

implies that

$$S^\alpha(i)\sigma(i + 1^\alpha; T(t_0)) = y^\alpha(i + 1^\alpha; T(t_0)) = f^\alpha(i, \sigma(i; T(t_0)), u(i; T(t_0)));$$

i.e., the pair $(\sigma(\cdot; T(t_0)), u(\cdot; T(t_0)))$ is admissible. Moreover, since

$$\Theta_0(t_0, T(t_0)) = 0,$$

it follows that

$$I(\sigma(\cdot; T(t_0)), u(\cdot; T(t_0))) - I(x^*, u^*) \equiv f_0(\sigma(k; T(t_0))) - f_0(x^*(k)) = -t_0 < 0.$$

In such a case, the pair (x^*, u^*) cannot be optimal. The proof is complete.

Remarks. It is well known that some form of convexity hypothesis is needed to establish a global minimum principle for discrete-time systems (Ref. 20), except for a very restricted class of problems (Ref. 37). Simple counterexamples to Theorem 3.1 can be found if the convexity hypothesis (H3) is not satisfied (Ref. 15).

The notion of a derived cone has been used in the present paper as a conical approximation to the state constraint sets. This notion was discussed by Hestenes (Ref. 38) and was used by Hautus (Ref. 25) in developing a general discrete-time maximum principle. The definition used in the present paper differs slightly from those used by the above-named authors (which also differ slightly from each other); the minor differences are discussed in Ref. 15. In a recent notable paper (Ref. 39), Martin *et al.* give an exhaustive comparison of the many types of convex conical approximations that have appeared in the literature.

There is a natural correspondence between the discrete multiparameter minimum principle of the present paper and the minimum principle developed by Cesari in Ref. 12 for partial differential equations in Dieudonné-Rashevsky form. A detailed comparison is given in Ref. 15.

5. Applications

Discrete multiparameter dynamic models, whether formulated *ab initio* as a discrete representation or obtained as a discretization of a continuum representation, frequently take the form of second-order or higher-order partial difference equations. These equations convey an insight into the underlying physical process being modeled, which may be obscured or destroyed by transforming the model into another representation purely for mathematical convenience. For example, one may produce some equivalent *canonical set* of first-order partial difference equations, or perhaps select an ordering of the independent discrete variables leading to an equivalent discrete-time system. The latter approach is often particularly undesirable; indeed, this fact has helped to motivate the development of discrete multiparameter linear system theory (Ref. 40).

As mentioned previously, most discrete multiparameter models encountered in applications can be expressed in the general form proposed in Section 2. The minimum principle of Section 3 can thus be interpreted in terms of a physically motivated higher-order discrete model. Quite general optimization problems using a discrete representation associated with hyperbolic (causal), parabolic (semicausal), and elliptic (noncausal) partial differential equation models have been treated in this manner (Ref. 15).

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