NONLINEAR PROGRAMMING AND THE MAXIMUM PRINCIPLE FOR DISCRETE TIME OPTIMAL CONTROL PROBLEMS

by

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ABSTRACT

Results in nonlinear programming are used to prove a generalized version of the maximum principle for fixed-time discrete optimal control problems. Proofs are based upon the implicit function theorem and a theorem of the alternative for systems of linear inequalities over a convex set; they do not, as in the past, require Brouwer's fixed-point theorem.
Introduction

It is becoming apparent that optimal control theory and nonlinear programming are highly related. This point of view is illustrated by the recent text of Canon, Cullum and Polak [1]. It seems yet to be fully exploited, however.

Our purpose here is to utilize results from nonlinear programming to establish a generalized version of the maximum principle for fixed-time discrete optimal control problems. These problems have been treated in various degrees of generality by a number of researchers [1], [3], [4], [5], [9], [10], [11]. We shall consider an extension of a version originally discussed by Halkin [3] and later generalized by Holtzman [4] and Cannon, Cullum and Polak [1]. Thus, we will not require convexity of the set of admissible controls (though a certain convexity condition will be imposed) nor do we require differentiability with respect to the control variables. We also admit inequality constraints on the state variables as well as equality constraints on the initial and terminal state vectors.

For motivation, we might note that linear optimal problems trivially fit out framework. In addition, Halkin [3] has shown discrete approximations to continuous problems satisfy the hypothesis that we shall impose.

The nonlinear programming approach that we adopt should be contrasted with previous approaches. We do not require intricate arguments based upon Brouwer's fixed point theorem, nor do we rely upon canonical approximations [1]. Instead, we utilize two basic results from nonlinear programming stated as Lemma's 1 and 2 below. The first result is a direct consequence of the implicit function theorem and a well-known lemma (see [7]) due to Motzkin. It was established by Mangasarian and Fromowitz [7], [8]. A short
proof is given in [5]. Though the result is stated as a "maximum principle" in [7], it has not been previously used in the present optimal control context, but rather as a direct extension of the classical Fritz John theorem of nonlinear programming applied with differentiability requirements that we do not enforce.

The second result is essentially a special case of a theorem of Fan, Glicksburg and Hoffman [2], [7, p.63]. The version that we give can be easily established via an elementary separating hyperplane argument.

Before stating these results, let us set some notation. $\mathbb{R}$ denotes the real numbers, $\mathbb{R}^n$ $n$-dimensional real space (with the usual topology). Subscripting denotes distinct vectors and superscripting vector components. This same convention will be applied to functions in the sense that if $g: \mathbb{R}^n \rightarrow \mathbb{R}^m$, then $g^i$ denotes the $i$th coordinate function, $g^i: \mathbb{R}^n \rightarrow \mathbb{R}$.

$\nabla g(\bar{x})$ denotes the matrix $\left( h_{ij} \right) = \left( \frac{\partial g^i(\bar{x})}{\partial x^j} \right)_{x=\bar{x}}$. Similarly, if $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is a function, $\nabla f(\bar{x})$ is the gradient vector evaluated at $x=\bar{x}$, i.e.,

$$\left( \frac{\partial f(\bar{x})}{\partial x^1}, \ldots, \frac{\partial f(\bar{x})}{\partial x^n} \right)_{x=\bar{x}}.$$

Finally, if $\alpha, \nu \in \mathbb{R}^n$ and $A$ is a real valued matrix, then $\alpha \nu$ denotes inner product as does $\alpha A$ and $A \nu$, i.e., $(A \nu)^i = \sum_{j=1}^{n} A_{ij} \nu_j$. Also, vector equalities and inequalities hold componentwise.

Lemma 1 (First Linearization Lemma): Let $C \subseteq \mathbb{R}^n$ be a convex set with a non-empty interior and let $f: \mathbb{R}^n \rightarrow \mathbb{R}$, $\phi: \mathbb{R}^n \rightarrow \mathbb{R}^m_1$, $\xi: \mathbb{R}^n \rightarrow \mathbb{R}^m_2$ be given functions.

Let $\bar{x}$ solve \begin{align*}
\max f(x) \\
\text{subject to } \phi(x) = 0 \\
\xi(x) \leq 0 \\
x \in C
\end{align*}.
Assume that \( \phi(x) \) is continuously differentiable in a neighborhood of \( \bar{x} \) and that \( f(x) \) and \( \xi(x) \) are differentiable at \( \bar{x} \). Then, if the vectors 
\[
\{\nabla \phi^j(\bar{x}) : j=1,\ldots,m_1\}
\]
are linearly independent, the system
\[
\begin{align*}
\nabla f(\bar{x})(\Delta x) &> 0 \\
\nabla \phi(\bar{x})(\Delta x) &= 0 \\
\nabla \xi^k(\bar{x})(\Delta x) &< 0 \quad \text{for } i\in\{1 \leq j \leq m_2 : \xi^j(\bar{x})=0\}
\end{align*}
\]
\( \bar{x} + \Delta x \in \{\text{interior of } C\} \)

has no solution in the variables \((\Delta x)^k : k=1,\ldots,n\).

**Lemma 2:** Let \( C \subseteq \mathbb{R}^n \) be a convex set and let \( A \) and \( B \) be \( m_1 \times n \) and \( m_2 \times n \) real valued matrices. If the system
\[
\begin{align*}
Ax &= 0 \\
Bx &< 0 \\
x &\in C
\end{align*}
\]
has no solution, then there exist \( \sigma \in \mathbb{R}^{m_1} \), \( \alpha \in \mathbb{R}^{m_2} \), \( \alpha \leq 0 \), \( (\sigma,\alpha) \neq 0 \) such that
\[
\sigma(\Delta x) + \alpha(Bx) \leq 0 \quad \text{for all } x \in C. \tag{1}
\]

For convenience, let us record the following special case of this result to be used later.

**Remark 1:** Let \( \bar{C} \subseteq \mathbb{R}^m \) be a convex set; let \( A_1, A_2, B_1 \) and \( B_2 \) be respectively \( m_1 \times N \), \( m_1 \times M \), \( m_3 \times N \), and \( m_4 \times M \) real valued matrices. Then, by Lemma 2 if the system
\[
\begin{align*}
A_1 v + A_2 w + \rho &= 0 \\
B_1 v &< 0 \\
B_2 w &< 0 \\
\rho &\in \bar{C}
\end{align*}
\]
has no solution, then there exist \( \sigma \in \mathbb{R}^{m_1}, \psi \in \mathbb{R}^{m_3}, \mu \in \mathbb{R}^{m_4}, \psi < 0, \mu < 0, \)
\((\sigma, \psi, \mu) \neq 0\) satisfying

\[
\sigma A_1 + \psi B_1 = 0
\]
\[
\sigma A_2 + \mu B_2 = 0
\]
\[
\sigma \rho \leq 0 \text{ for all } \rho \in C.
\]

If there are no \(B_1\) and/or \(B_2\) constraints, we may eliminate \(\psi\) and/or \(\mu\) above. To obtain the equality constraints above, simply take positive and negative unit vectors for \(v\) and \(w\) in (*).
I. Problem Statement

Let $\Omega_0, \Omega_1, \ldots, \Omega_{T-1}$ be given subsets of $\mathbb{R}^r$. Assume that each of the following functions is given:

- $f_t: \mathbb{R}^n \times \Omega_t \rightarrow \mathbb{R}^n$ for $t=0,1,\ldots,T-1$
- $h: \mathbb{R}^n \rightarrow \mathbb{R}$
- $g_0: \mathbb{R}^n \rightarrow \mathbb{R}$
- $g: \mathbb{R}^n \rightarrow \mathbb{R}^m$
- $q_t: \mathbb{R}^n \rightarrow \mathbb{R}^m_t$ for $t=0,1,\ldots,T$.

Problem: Determine state vectors $x_0, x_1, \ldots, x_T \in \mathbb{R}^n$ and control vectors $u_0, u_1, \ldots, u_{T-1} \in \mathbb{R}^r$ to

\[
\max g_0(x_T)
\]

subject to

\[
h(x_0) = 0
\]
\[
x_{t+1} - x_t = f_t(x_t, u_t) \quad (t=0,1,\ldots,T-1)
\]
\[
g(x_T) = 0
\]
\[
q_t(x_t) \leq 0 \quad (t=0,1,\ldots,T)
\]
\[
u_t \in \Omega_t \quad (t=0,\ldots,T-1)
\]

Suppose that $\bar{x}_0, \bar{x}_1, \ldots, \bar{x}_T$ and $\bar{u}_0, \bar{u}_1, \ldots, \bar{u}_{T-1}$ solve $P$. Some typical hypotheses for the problem are:

H1) for every $u \in \Omega_t$, the vector valued function $f_t(x, u)$ is continuously differentiable with respect to $x$ in a neighborhood of $x=\bar{x}_t$ for $t=0,1,\ldots,T-1$.

H2) for every $x \in \mathbb{R}^n$, the set $\{f_t(x, u) : u \in \Omega_t\}$ is a convex subset of $\mathbb{R}^n$ for $t=0,1,\ldots,T-1$.

H3) the vector valued functions $h(x)$ and $g(x)$ are each continuously differentiable in neighborhoods of $x=\bar{x}_0$ and $x=\bar{x}_T$, respectively. $g_0(x)$ and the vector valued functions $q_t(x)$ for $t=0,1,\ldots,T$ are differentiable at respectively $x=\bar{x}_0, \bar{x}_T, \ldots, \bar{x}_T$. 

H4) the vectors $Vh^{j}(x) (j=1, \ldots, m)$ are linearly independent as are the vectors $Vg^{j}(x) (j=1, \ldots, m)$

H5) For each $t=0, 1, \ldots, T$ either no $q_{t}(x_{t}) = 0$ $(i=1, \ldots, m)$ or there is no nonzero

$$\lambda_{t} \in \mathbb{R}^{m}, \lambda_{t} \leq 0$$

satisfying

$$\lambda_{t} Vq_{t}(x_{t}) = 0$$

$$\lambda_{t} q_{t}(x_{t}) = 0.$$ 

(Note that if for each fixed $t$, the vectors \{v_{t}^{i}(x_{t}); q_{t}^{i}(x_{t})=0\} are linearly independent, then (H5) holds.)

The discrete maximum principle for $P$ states that there exists (adjoint) vectors $p_{0}, p_{1}, \ldots, p_{T} \in \mathbb{R}^{n}$, and (multiplier) vectors $\lambda_{t} \in \mathbb{R}^{m}$, $\lambda_{t} \leq 0$ $t=0, 1, \ldots, T-1$,

vectors $\alpha \in \mathbb{R}^{n}$, $\beta \in \mathbb{R}^{m}$ and a scalar $\beta > 0$ satisfying:

0) Not all the quantities $p_{0}, p_{1}, \ldots, p_{T}, \alpha, \beta$ and $\beta_{0}$ are zero.

1) Hamiltonian maximization:

for all $u \in \Omega$, $p_{t+1} f_{t}(x_{t}, u) \geq p_{t+1} f_{t}(x_{t}, u)$ $(t=0, 1, \ldots, T-1)$

2) Adjoint equations:

$$p_{t} - p_{t+1} = p_{t+1} V_{x} f_{t}(x_{t}, u_{t}) + \lambda_{t} V_{q_{t}}(x_{t})$$

$(t=0, 1, \ldots, T-1)$

3) Transversality conditions:

$$p_{0} = \alpha V_{h}(x_{0})$$

$$p_{T} = \beta V_{g}(x_{T})$$

$+ \beta_{0} V_{g_{0}}(x_{T}) + \lambda_{T} V_{q_{T}}(x_{T})$

4) Complementary slackness: $\lambda_{t} q_{t}(x_{t}) = 0$ $(t=0, 1, \ldots, T)$

Our purpose will be to prove an extended version of the fact that (H1)-(H5) imply this discrete maximum principle. We will use the first linearization lemma and Remark 1 as basic tools.

Towards this end, it will be convenient to consider a generalized version of $P$. Let $U$ be a given subset of $\mathbb{R}^{L}$ and suppose that each of the following functions are given:

$$F : \mathbb{R}^{N} \rightarrow \mathbb{R}$$

$$H : \mathbb{R}^{N} \times \mathbb{R}^{M} \times U \rightarrow \mathbb{R}^{r}$$

$$G : \mathbb{R}^{N} \rightarrow \mathbb{R}^{m}$$

$$Q : \mathbb{R}^{N} \rightarrow \mathbb{R}^{s_{1}}$$

$$\bar{Q} : \mathbb{R}^{M} \rightarrow \mathbb{R}^{s_{2}}$$

$v_{t}^{i}(x_{t}, u_{t})$ denotes the matrix $Vd(x_{t})$ where $d(x)=f_{t}(x, u_{t})$. This same convention applies to any function of two or more vectors.
The new problem is:

Determine \( y \in \mathbb{R}^N, z \in \mathbb{R}^M \) and \( u \in \mathbb{R}^L \) to

max \( F(y) \)

subject to \( G(y) = 0 \)

\( Q(y) \preceq 0 \)

\( \bar{Q}(z) \preceq 0 \)

\( H(y,z,u) = 0 \)

\( u \in U \).

Assume that \( \bar{y}, \bar{z}, \bar{u} \) solve (P'). We shall impose the following analogs of (H1)-(H5).

\( \overline{H1} \) For every \( u \in U \) the vector valued function \( H(y,z,u) \) is continuously differentiable with respect to \( (y, z) \) in a neighborhood of \( (y,z) = (\bar{y}, \bar{z}) \).

\( \overline{H2} \) For every \( y \in \mathbb{R}^N, z \in \mathbb{R}^M \), the set \( \{H(y,z,u) : u \in U\} \) is a convex subset of \( \mathbb{R}^r \).

\( \overline{H3} \) \( G(y) \) is continuously differentiable in a neighborhood of \( y = \bar{y} \). \( F(y) \) and \( Q(y) \) are differentiable at \( y = \bar{y} \) and \( \bar{Q}(z) \) is differentiable at \( z = \bar{z} \).

\( \overline{H4} \) The vectors \( vG^j(y) \) \( (j = 1, \ldots, m) \) are linearly independent.

\( \overline{H5} \) Let \( A = \{i \in \{1, \ldots, s_1\} : Q^i(\bar{y}) = 0\} \)

\( \bar{A} = \{i \in \{1, \ldots, s_2\} : \bar{Q}^i(\bar{z}) = 0\} \) and

\( \bar{s}_1 = |A|, \bar{s}_2 = |\bar{A}| \) \( (\cdot) \) is cardinality.

Also, let \( Q^A : \mathbb{R}^N \rightarrow \mathbb{R}^{s_1} \) denote the vector valued function defined by restricting the \( Q^i \) \( (i = 1, \ldots, s_1) \) to \( A \) and similarly define \( \bar{Q}^\bar{A} \) (z). Then

\( A = \emptyset \) or there is no non-zero \( \mu^A \in \mathbb{R}^{s_1} \), \( \mu^A \preceq 0 \) with \( \mu^A Q^A(\bar{y}) = 0 \) and

\( \bar{A} = \emptyset \) or there is no non-zero \( \bar{\mu}^\bar{A} \in \mathbb{R}^{s_2} \), \( \bar{\mu}^\bar{A} \preceq 0 \) with \( \bar{\mu}^\bar{A} \bar{Q}^\bar{A}(\bar{z}) = 0 \).

Note that problem P corresponds to the case where \( z = \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_{T-1} \end{pmatrix} \), \( u = \begin{pmatrix} u_0 \\ u_1 \\ \vdots \\ u_{T-1} \end{pmatrix} \), \( y = x_T \).
\[
U = \Omega_0 x_0 x_1 \ldots x_{T-1}
\]
\[
F(y) = g_0(x_T)
\]
\[
G(y) = g_T(x_T)
\]
\[
H(y, z, u) = 
\begin{pmatrix}
  h(x_0) & x_0 f_0(x_0, u_0) & -x_1 \\
  x_1 f_1(x_1, u_1) & -x_2 & \ldots & -x_{T-2} \\
  \vdots & \ddots & \ddots & \ddots \\
  x_{T-1} f_{T-1}(x_{T-1}, u_{T-1}) & \ldots & x_T
\end{pmatrix}
\]
\[
Q(z) = 
\begin{pmatrix}
  q_0(x_0) \\
  q_1(x_1) \\
  \vdots \\
  q_{T-1}(x_{T-1})
\end{pmatrix}
\]
\[
Q(y) = q_T(x_T)
\]

Note that
\[
[V H(y, z, u), V_y H(y, z, u)] = 
\begin{pmatrix}
  \nabla h(x_0) & I + \nabla f_0(x_0, u_0) & -I \\
  I + \nabla f_1(x_1, u_1) & \ldots & -I \\
  \vdots & \ddots & \ddots & \ddots \\
  I + \nabla f_{T-1}(x_{T-1}, u_{T-1}) & \ldots & I + \nabla f_{T-1}(x_{T-1}, u_{T-1}) & -I
\end{pmatrix}
\]

With these associations (H1)-(H5) are direct consequences of (H1)-(H5).
II. Generalized Discrete Maximum Principle

The proof of a generalized discrete maximum principle that we are to give will be based in part upon an elementary fact concerning matrix valued functions. Let $U$ be a given set and suppose that for each $u \in U$, $\psi(u)$ is a real valued matrix. We will call $\psi$ linearly independent over $U$ if there is no $\pi \in \mathbb{R}^{d_1}$, $\pi \neq 0$ such that $\pi \psi(u) = 0$ for all $u \in U$.

**Remark 2:** $\psi$ is linearly independent over $U$ if and only if there are $u_1, \ldots, u_D \in U$ such that the matrix

$$M = [\psi(u_1), \psi(u_2), \ldots, \psi(u_D)]$$

has full row rank.

**Proof:**

If $\pi \in \mathbb{R}^{d_1}$, $\pi \neq 0$ and $\pi \psi(u) = 0$ for all $u \in U$, then $\pi M = 0$ for any choice of $u_1, u_2, \ldots, u_D \in U$. Thus no $M$ above has full row rank.

Conversely, suppose that $\psi$ is linearly independent over $U$ and that $M$ is given. Let rows $r_1, \ldots, r_k$, $k < d_1$ be linearly independent rows of $M$ spanning its row space. Then any other row $r_{k+1}$ of $M$ can be expressed uniquely via $\pi M = 0$ with $\pi_{k+1} = 1$, $\pi_i = 0$ for $i \neq 1, 2, \ldots, k$. By the linear independence of $\psi$, there is a $u_{D+1}$ such that $\pi \psi(u_{D+1}) \neq 0$. Thus, rows $r_1, \ldots, r_k, r_{k+1}$ of $[\psi(u_1), \ldots, \psi(u_D), \psi(u_{D+1})]$ are linearly independent. This matrix either has full row rank or we may continue adding $\psi(u_j)$ as above until the resulting matrix does.

**Remark 3:** Suppose that $\psi$ above is given by $\psi(u) = \begin{pmatrix} A & \phi(u) \end{pmatrix}$ for fixed real matrices $A$ and $B$ where $\phi(u)$ is a real valued matrix. Then $\psi$ is linearly independent if and only if there are $u_1, \ldots, u_D \in U$ such that

$$M' = \begin{pmatrix} A & \phi(u_1) & \ldots & \phi(u_D) \\ B & \phi(u_2) & \ldots & B \phi(u_D) \end{pmatrix}$$

has full row rank.

**Proof:** The row ranks of $\begin{pmatrix} A & \phi(u_1) & \ldots & \phi(u_D) \\ B & \phi(u_2) & \ldots & B \phi(u_D) \end{pmatrix}$ and $M'$ are the same.

Using these results, the first linearization lemma and Remark 1, we now show:
Theorem 1 (Generalized Discrete Maximum Principle)

Let \( y = \bar{y}, x = \bar{x} \) and \( u = \bar{u} \) solve control problem \( P' \); assume \( \overline{H1-H3} \) and \( \overline{H5} \). Then there exist a real number \( \beta_0 \geq 0 \) and vectors \( \bar{\beta} \in \mathbb{R}^m, \bar{\gamma} \in \mathbb{R}^r, \mu \in \mathbb{R}^s_1, \bar{\mu} \in \mathbb{R}^s_2, \delta \in \mathbb{R}^n \) with \( \mu \leq 0, \bar{\mu} \leq 0, (\beta_0, \beta, \sigma) \neq 0 \), satisfying

(a) \(-\delta = \beta_0 \nabla F(\bar{y}) + \bar{\beta} \nabla G(\bar{y}) + \mu \nabla Q(\bar{y})\)

(b) \(\delta = \sigma \nabla_y H(\bar{y}, \bar{z}, \bar{u})\)
\(0 = \sigma \nabla_z H(\bar{y}, \bar{z}, \bar{u}) + \bar{\mu} \nabla \bar{Q}(\bar{z})\)

(c) \(\mu Q(\bar{y}) = 0, \quad \bar{\mu} Q(\bar{z}) = 0\)

(d) \(\sigma[H(\bar{y}, \bar{z}, \bar{u}) - H(\bar{y}, \bar{z}, u)] \geq 0 \) for all \( u \in U \).

Proof: If the matrix valued function
\[
\psi(u) = \begin{pmatrix}
\nabla G(\bar{y}) \\
\nabla_y H(\bar{y}, \bar{z}, \bar{u}) \\
\nabla_z H(\bar{y}, \bar{z}, \bar{u}) \\
H(\bar{y}, \bar{z}, u)
\end{pmatrix}
\]
is not linearly independent over \( U \), then there are \( \beta \in \mathbb{R}^m, \sigma \in \mathbb{R}^r, (\beta, \sigma) \neq 0 \) such that
\[
\delta = -\beta \nabla G(\bar{y}) = \sigma \nabla_y H(\bar{y}, \bar{z}, \bar{u})
\]
\(\sigma \nabla_z H(\bar{y}, \bar{z}, \bar{u}) = 0\)
\(\sigma H(\bar{y}, \bar{z}, u) = 0\)
for all \( u \in U \).

Note that \( \sigma \neq 0 \) since by \( \overline{H4} \) the rows of \( \nabla G(\bar{y}) \) are linearly independent.

Taking \( \beta_0, \mu, \bar{\mu} \) all zero and incorporating \( H(\bar{y}, \bar{z}, \bar{u}) = 0 \) in \( \sigma H(\bar{y}, \bar{z}, u) = 0 \), (a)-(d) is satisfied.

On the other hand, if \( \psi \) is linearly independent over \( U \), then by Remark 3, there are \( u_1, \ldots, u_D \in U \) such that
\[
M' = \begin{bmatrix}
\nabla G(\bar{y}) \\
\nabla_y H(\bar{y}, \bar{z}, \bar{u}) \\
\nabla_z H(\bar{y}, \bar{z}, \bar{u}) \\
H(\bar{y}, \bar{z}, u_1) \\
\vdots \\
H(\bar{y}, \bar{z}, u_D) \\
H(\bar{y}, \bar{z}, u)
\end{bmatrix}
\]
has full row rank. This will provide the hypothesis for an application of Lemma 1.
Let \( u \in U \) be fixed. By \( H2 \), given any \( \theta_j \geq 0 \) \( (j=1, \ldots, D+1) \), \( \sum_{j=1}^{D+1} \theta_j \leq 1 \), \( y \in \mathbb{R}^N \), \( z \in \mathbb{R}^M \), there is a \( u(\theta_1, \ldots, \theta_{D+1}, y, z) \in U \) such that

\[
H(y, z, u(\theta_1, \ldots, \theta_{D+1}, y, z)) = (1 - \sum_{j=1}^{D} \theta_j)H(y, z, \bar{u}) + \sum_{j=1}^{D} \theta_j H(y, z, u_j) + \theta_{D+1} H(y, z, u) \geq 0
\]

Consequently, \( y = \bar{y}, z = \bar{z}, \theta_j = \delta_j \neq 0 \) \( (j=1, \ldots, D+1) \) must solve the problem (in the variables \( y, z, \theta_1, \ldots, \theta_{D+1} \)):

\[
\max \ F(y) \quad \text{subject to} \quad G(y) \leq 0, \quad Q^A(y) \leq 0 \quad (\bar{P})
\]

\[
Q^A(z) \leq 0, \quad (D+1) \sum_{j=1}^{D} \theta_j H(y, z, \bar{u}) + \sum_{j=1}^{D} \theta_j H(y, z, u_j) + \theta_{D+1} H(y, z, u) = 0
\]

\[
(y, z, \theta_1, \ldots, \theta_{D+1}) \in \mathcal{C} = \{(y, z, \theta_1, \ldots, \theta_{D+1}) : y \in \mathbb{R}^N, z \in \mathbb{R}^M, \sum_{j=1}^{D+1} \theta_j \leq 1, \theta_j \geq 0 \ (j=1, \ldots, D+1)\}
\]

Utilizing \( H1 \) and \( H3 \), we see that this problem satisfies the hypothesis of the first linearization lemma with the association \( x = (y, z, \theta_1, \ldots, \theta_{D+1}) \) and the obvious associations for \( \phi \) and \( \xi \). Note that \( \nabla \phi(\bar{x}) = M' \) above which has full row rank. By that lemma, the following system has no solution:

\[
\nabla F(\bar{y})v > 0 \quad (1)
\]
\[
\nabla G(\bar{y})v = 0 \quad (2)
\]
\[
\nabla Q^A(\bar{y})v < 0 \quad (3)
\]
\[
\nabla Q^A(\bar{z})w < 0 \quad (4)
\]
\[
\nabla Y(\bar{y}, \bar{z}, \bar{u})v + \nabla Z(\bar{y}, \bar{z}, \bar{u})w + \sum_{j=1}^{D} H(\bar{y}, \bar{z}, u_j) \delta_j + H(\bar{y}, \bar{z}, u) \delta_{D+1} = 0 \quad (5)
\]
\[
\sum_{j=1}^{D+1} \delta_j < 1, \delta_j > 0 \ (j=1, \ldots, D+1) \quad (6)
\]

Observe that we have used \( H(\bar{y}, \bar{z}, \bar{u}) = 0 \) here.

Note that by scaling \( (1)-(6) \) has no solution if and only if \( (1)-(5) \) and

\[
\theta_j > 0 \ (j=1, \ldots, D+1) \quad (6')
\]
has no solution. Also note that we may write (5) as

\[ \nabla_y H(\bar{y}, \bar{z}, \bar{u}) v + \nabla_z H(\bar{y}, \bar{z}, \bar{u}) w + \sum_{j=1}^{D} H(\bar{y}, \bar{z}, u_j) \hat{\theta}_j + \rho = 0 \]  

(5')

where \( \rho = H(\bar{y}, \bar{z}, u) \hat{\theta}_{D+1} \)

Consequently (1)-(4), (5') and (6') have no solution for any

\[ \rho \in S = \{ H(\bar{y}, \bar{z}, u) \hat{\theta}_{D+1} \text{ for some } u \in U, \hat{\theta}_{D+1} > 0 \} \]

Observe that \( S \) is a convex set, since if \( \rho_1, \rho_2 \in S \)

then \( \rho_1 = H(\bar{y}, \bar{z}, u) \hat{\theta}_{D+1} \)

and \( \rho_2 = H(\bar{y}, \bar{z}, u') \hat{\theta'}_{D+1} \) for some \( u, u' \in U, \hat{\theta}_{D+1} > 0, \hat{\theta'}_{D+1} > 0 \)

But then for any \( \lambda \in [0,1] \),

\[ \lambda \rho_1 + (1-\lambda) \rho_2 = [\lambda \hat{\theta}_{D+1} + (1-\lambda) \hat{\theta'}_{D+1}] \{ \frac{\lambda H(\bar{y}, \bar{z}, u)}{\lambda \hat{\theta}_{D+1} + (1-\lambda) \hat{\theta'}_{D+1}} \}

By \( H_2 \), the bracketed term equals \( H(\bar{y}, \bar{z}, \bar{u}) \) for some \( \bar{u} \in U \); thus

\[ \lambda \rho_1 + (1-\lambda) \rho_2 \in S \]

Consequently we may apply Remark 1 to system (1)-(4), (5') and (6') over \( S \), i.e., there is a scalar \( \beta_0 \geq 0 \) and vectors \( \beta \in \mathbb{R}^m \), \( \sigma \in \mathbb{R}^r \), \( \mu^A \in \mathbb{R}^{g_1} \), \( \mu^A \in \mathbb{R}^{g_2} \) with

\[ \mu^A \leq 0, \mu^A \leq 0 \text{ and } (\beta_0, \beta, \sigma, \mu^A, \mu^A) \neq 0 \]

satisfying

\[ \beta_0 \nabla F(\bar{y}) + \beta \nabla G(\bar{y}) + \mu^A \nabla \bar{A}(\bar{y}) + \sigma \nabla H(\bar{y}, \bar{z}, \bar{u}) = 0 \]

\[ \sigma \nabla H(\bar{y}, \bar{z}, \bar{u}) + \mu^A \nabla \bar{A}(\bar{z}) = 0 \]

\[ \sigma H(\bar{y}, \bar{z}, u_j) \hat{\theta}_j \leq 0 \text{ for all } \hat{\theta}_j > 0 \text{ (} j=1, \ldots, D \text{)} \]

\[ \sigma H(\bar{y}, \bar{z}, u) \hat{\theta}_{D+1} \leq 0 \text{ for all } \hat{\theta}_{D+1} > 0 \text{ and } u \in U. \]
If \( A \neq \emptyset \), then \( \sigma \neq 0 \) by \( \overline{H}_5 \). If \( A = \emptyset \) and \( \sigma = 0 \), then \((\beta_0, \beta) \neq 0\) by \( \overline{H}_5 \). Thus \((\beta_0, \beta, \sigma) \neq 0\). Also, the last consequence of Remark 1 above implies that \( \omega H(\overline{y}, \overline{z}, \overline{u}) \leq 0 \) for all \( u \in U \). Incorporating \( H(\overline{y}, \overline{z}, \overline{u}) = 0 \) in this last statement, defining \( \delta = \sigma V \overline{y} H(\overline{y}, \overline{z}, \overline{u}) \) and letting \( \bar{u} = 0 \) for \( \bar{u} \in A \), \( \bar{u} = 0 \) for \( \bar{u} \in A \), \((a)-(d)\) is satisfied.

Remark 4: If the vectors \( V G^j(\overline{y}) \) \((j=1, \ldots, m)\) are linearly independent, then theorem 1 holds by taking \( \beta V G(\overline{y}) = 0 \), \( \beta \neq 0 \) and \((\beta_0, \sigma, \mu, \bar{u}, \delta) = 0\). Thus \( H_4 \) will rule out this trivial case.

Corollary 1.1 (Discrete maximum principle) Let \( u_t = \overline{u}_t \) \((t=0, \ldots, t-1)\) and \( x_t = \overline{x}_t \) \((t=0, \ldots, T)\) solve control problem \( P \); assume \( H_1 \) through \( H_5 \). Then the discrete maximum principle holds.

Proof: We previously showed that \( P \) is a special case of \( P' \) and that \( \overline{H}_1-\overline{H}_5 \) hold when we make this association. Let \( \sigma, \mu, \bar{u} \) and \( \delta \) from Theorem 1 be given by:

\[
\sigma = (-\alpha_0, p_0, p_1, \ldots, p_{T-1})
\]
\[
\mu = \lambda_T
\]
\[
\bar{u} = (\lambda_1, \ldots, \lambda_{T-1})
\]
\[
\delta = -p_T
\]

In these terms, consequences \((a)-(d)\) of Theorem 1 are translated into the following terms for \( P \).

\[
(a) \quad p_T = \beta_0 \nabla g_0(\overline{x}_T) + \beta g(\overline{x}_T) + \lambda_T q_T(\overline{x}_T)
\]

\[
(b) \quad -p_T = -p_T - \alpha \nabla h(\overline{x}_0) + p_1[I+ \nabla f_0(\overline{x}_0, \overline{u}_0)] + \lambda_0 q_0(\overline{x}_0) = 0
\]

\[
- p_T + p_{t+1}[I+ \nabla f_t(\overline{x}_t, \overline{u}_t)] + \lambda_{t+1} q_t(\overline{x}_t) = 0 \quad (t=1, \ldots, t-1)
\]

\[
(c) \quad \lambda_{T-1} q_T(\overline{x}_T) = 0, \quad \sum_{t=0}^{T-1} \lambda_t q_t(\overline{x}_t) = 0
\]

\[
(d) \quad \sum_{t=0}^{T-1} p_{t+1} f_t(\overline{x}_t, \overline{u}_t) \geq \sum_{t=0}^{T-1} p_{t+1} f_t(\overline{x}_t, \overline{u}_t) \quad \text{for all}
\]
\[
(u_0, \ldots, u_{T-1}) \in \Omega \times \Omega \times \cdots \times \Omega
\]
Defining \( p_0 = \alpha v h(x_0) \), (a) and (b) become the adjoint equations and transversality conditions. Since \( q_t(x_t) \leq 0 \) and \( \lambda_t \leq 0 \) (c) implies the complementary conditions \( \lambda_t q_t(x_t) = 0 \) \((t=0, \ldots, T-1)\). Successively taking \( u_j = u_j \) \((j \neq t)\) for \( t=0,1, \ldots, T-1 \), (d) implies Hamiltonian maximization for each \( t=0, \ldots, T-1 \). Finally, \( p_0, p_1, \ldots, p_T, \alpha, \beta \) and \( \beta_0 \) are not all zero from Theorem 1 and the above associations.

**Remark 5:**

(i) As in remark 4, \( H4 \) rules out the trivial case where \( \beta v g(y) = 0 \) and \( \beta \) is the only non-zero multiplier or adjoint variable. Similarly \( H4 \) rules out the analogous case where \( \alpha v h(y) = 0 \), \( \alpha \neq 0 \).

(ii) In the case of no inequality constraints on state variable, non-singularity of \([I+V x_t(t,t)](t=0, \ldots, t-1)\) implies that \( p_t \neq 0 \) \((t=0,1, \ldots, T-1)\). If in addition the vectors \( v g(x_t), v g^1(x_t), \ldots, v g^m(x_t) \) are linearly independent then \( p_T \neq 0 \).

**III. An extension**

Holtzman [4] extended the discrete maximum principle by allowing \( g_0 \) to be a function of \( u \) as well as \( x_t \) and introducing the concept of directional convexity. We show here that an analogous assumption permits our proof of the generalized maximum principle to be easily extended. When only \( F \) and \( H \) below are functions of \( u \), assumption \( H2' \) is equivalent to Holtzman's definition of directional convexity (see [1]).

Suppose that \( y=\tilde{y}, z=\tilde{z}, u=\tilde{u} \) solves the control problem:

\[
\begin{align*}
\text{max } & F(y,u) \\
\text{subject to } & G(y,u) \leq 0 \\
& Q(y,u) \leq 0 \\
& \bar{Q}(z,u) \leq 0 \\
& H(y,z,u) = 0 \\
& u \in U \\
\end{align*}
\]

\((P'')\)

Let \( H1' \) be the same as \( H1 \) and modify \( H2-H5 \) to:

\( H2' \) for every \( y \in R^N, z \in R^M, u \in U, u' \in U, \lambda \in [0,1] \) there is a \( \tilde{u} \in U \) such that
\[ F(y, \tilde{u}) \geq \lambda F(y, u) + (1-\lambda) F(y, u') \]
\[ Q(y, \tilde{u}) \leq \lambda Q(y, u) + (1-\lambda) Q(y, u') \]
\[ \tilde{Q}(z, \tilde{u}) \leq \lambda \tilde{Q}(z, u) + (1-\lambda) \tilde{Q}(z, u') \]
\[ G(y, \tilde{u}) = \lambda G(y, u) + (1-\lambda) G(y, u') \]
\[ H(y, z, \tilde{u}) = \lambda H(y, z, u) + (1-\lambda) H(y, z, u') \]

\( H3' \) For every u \in U, G(y, u) is continuously differentiable in a neighborhood of y=\tilde{y}, F(y, u) and Q(y, u) are differentiable at y=\tilde{y}, and \( \tilde{Q}(z, u) \) is differentiable at z=\tilde{z}.

\( H4' \) The vectors \( V G(y, y', z) \) \( (j=1, \ldots, m) \) are linearly independent.

\( H5' \) \( H5 \) with \( Q_i(y) \) and \( Q_i(y) \) replaced by \( Q_i(y, \tilde{u}) \) and \( \tilde{Q}_i(z, \tilde{u}) \) in the definitions of A and \( \tilde{A} \).

With these modifications Theorem 1 becomes:

**Theorem 1A:**

Let y=\tilde{y}, x=\tilde{x}, u=\tilde{u} solve control problem P'; assume \( H1'-H3' \) and \( H5' \). Then there exist a real number \( \beta > 0 \) and vectors \( \beta \in \mathbb{R}^m, \sigma \in \mathbb{R}^r, \mu \in \mathbb{R}^2 \), \( \alpha \in \mathbb{R}^2, \beta \in \mathbb{R}^r \) with \( \mu < 0, \beta < 0, (\beta, \sigma, \mu) \neq 0 \) satisfying

\[ (a') -\delta = \beta \nabla_y F(\tilde{y}, \tilde{u}) + \beta \nabla_y G(\tilde{y}, \tilde{u}) + \mu \nabla_y Q(\tilde{y}, \tilde{u}) \]
\[ (b') \delta = \sigma \nabla_y H(\tilde{y}, \tilde{z}, \tilde{u}) \]
\[ 0 = \sigma \nabla_z H(\tilde{y}, \tilde{z}, \tilde{u}) + \mu \nabla_z \tilde{Q}(\tilde{z}, \tilde{u}) \]
\[ (c') \mu Q(\tilde{y}, \tilde{u}) = 0, \quad \mu Q(\tilde{y}, \tilde{u}) = 0 \]
\[ (d') [\mathcal{L}(u) - \mathcal{L}(u)] \geq 0 \text{ for all } u \in U \]

where \( \mathcal{L}(u) = \beta F(\tilde{y}, u) + \beta G(\tilde{y}, u) + \mu Q(\tilde{y}, u) + \mu \tilde{Q}(\tilde{z}, u) + \sigma H(\tilde{y}, \tilde{z}, u) \)

**Proof:** The proof is analogous to that of Theorem 1 and will be omitted. We simply note that \( \psi(u) \) is now defined as

\[ \psi(u) = \begin{pmatrix} \nabla_y G(\tilde{y}, u) & G(\tilde{y}, u) \\ \nabla_y H(\tilde{y}, \tilde{z}, u) & \nabla_z H(\tilde{y}, \tilde{z}, u) \end{pmatrix} \]
Also, in problem \( P \), functions \( F, G, Q^A \) and \( Q^\tilde{A} \) are each replaced by analogs of the form for \( H \) in that problem. For example, \( F(y) \) is replaced by,

\[
\left( 1 - \sum_{j=1}^{D+1} \theta_j \right) F(y,u) + \sum_{j=1}^{D} \theta_j F(y,u_j) + \theta_{D+1} F(y,u)
\]

**Remark 6:** \( H^4' \) admits a statement analogous to Remark 4. In addition, the statement of Theorem 1A can be specialized to an extended version of problem \( P \), giving an extension of the discrete maximum principle. Details are omitted.
IV. A Separation Property

Let us consider the following linearized version of (P') about (\(\tilde{y}, \tilde{z}, \tilde{u}\)):

\[
\begin{align*}
\text{max } & \ V_F(\tilde{y})v \\
\text{subject to } & \ V_G(\tilde{y})v = 0 \quad \text{[}v = (y - \tilde{y}), w = (x - \tilde{x})\text{]} \\
\ & \ V_Q^A(\tilde{y})v < 0 \\
\ & \ V_Q^A(\tilde{z})w < 0 \\
\ & \ V_y H(\tilde{y}, \tilde{z}, \tilde{u})v + V_z H(\tilde{y}, \tilde{z}, \tilde{u})w + H(\tilde{y}, \tilde{z}, \tilde{u}) = 0 \\
\end{align*}
\]

This is a standard first order strict inequality approximation except for the linearized H equation. Here we have omitted \(G(y) = 0, Q^A(\tilde{y}) = 0, Q^A(\tilde{z}) = 0, H(\tilde{y}, \tilde{z}, \tilde{u}) = 0\) and the constant term \(F(\tilde{y})\).

Let \(S_1 = \{v \in \mathbb{R}^n: V_F(\tilde{y})v > 0, V_G(\tilde{y})v = 0, V_Q^A(\tilde{y})v < 0\}\) and

\(S_2 = \{v \in \mathbb{R}^n: V_Q^A(\tilde{z})w < 0, V_y H(\tilde{y}, \tilde{z}, \tilde{u})v + V_z H(\tilde{y}, \tilde{z}, \tilde{u})w + H(\tilde{y}, \tilde{z}, \tilde{u}) = 0\}\)

for some \(u \in U, w \in M\)

\(S_1\) and \(S_2\) each reflect half of the above linearized problem.

Halkin's [3] approach to the discrete maximum principle was to first prove that there is a hyperplane separating \(S_1\) and \(S_2\). He considered the case with no state constraints \(Q\) or \(\bar{Q}\) and with \(H\) having the associations for the control problem \(P\) that were given in section I. Since this separation property may be of independent interest, let us see how it results from the arguments of Theorem 1.

**Lemma 3 (Second Linearization Lemma):**

Let \(y = \tilde{y}, x = \tilde{x}\) and \(u = \tilde{u}\) solve control problem \(P'\); assume \(H_1-H_3\) and \(H_5\).

and that the matrix \([V_y H(\tilde{y}, \tilde{z}, \tilde{u}), V_z H(\tilde{y}, \tilde{z}, \tilde{u})]\) has full row rank.

Then \(S_1\) and \(S_2\) are separated by a hyperplane

**Proof:** For the most part, the proof is the same as the proof of Theorem 1. Consequently, we adopt the notation there and only sketch modifications to that proof.

If \(\psi(u)\) is not linearly independent over \(u\), then let \((\beta, \sigma) \neq 0\) and \(\delta\) be defined as before, that is
\[ \delta \equiv -\beta \nabla G(\overline{y}) = \sigma \nabla \chi(\overline{y}, \overline{z}, \overline{u}) \]
\[ \sigma \nabla \chi(\overline{y}, \overline{z}, \overline{u}) = 0 \]
\[ \sigma H(\overline{y}, \overline{z}, u) = 0 \text{ for all } u \in U \]

Then consider the hyperplane \( \{x : \delta x = 0\} \). Note that \( \delta \neq 0 \) since the rows of \( \nabla G(\overline{y}) \) are linearly independent as are those of 
\[ [\nabla \chi(\overline{y}, \overline{z}, \overline{u}), \nabla \chi(\overline{y}, \overline{z}, \overline{u})]. \]
For \( v \in S_1 \), \( \delta v = -\beta \nabla G(\overline{y})v = 0 \). For \( v \in S_2 \), there are \( w \in \mathbb{R}^M \) and \( u \in U \) so that \( \delta v = \sigma \nabla \chi(\overline{y}, \overline{z}, \overline{u})v = \sigma [\nabla \chi(\overline{y}, \overline{z}, \overline{u})v + \nabla \chi(\overline{y}, \overline{z}, \overline{u})w + H(\overline{y}, \overline{z}, u)] = 0 \).
Thus \( \{x : \delta x = 0\} \) separates \( S_1 \) and \( S_2 \).

If \( \psi(u) \) is linearly independent over \( u \), then let
\[ S_2^* = \{v \in \mathbb{R}^N : \nabla A(z)w < 0, \]
\[ \nabla \chi(\overline{y}, \overline{z}, \overline{u})v + \nabla \chi(\overline{y}, \overline{z}, \overline{u})w + \sum_{j=1}^{D} \nabla \chi(\overline{y}, \overline{z}, \overline{u}_j)\theta_j + H(\overline{y}, \overline{z}, \overline{u})\theta_{D+1} = 0 \]
\[ \sum_{j=1}^{D+1} \theta_j < 1, \quad \theta_j > 0, \quad (j = 1, \ldots, D+1) \text{ for some } u \in U, \ w \in \mathbb{R}^M \}

where \( u_1, \ldots, u_D \) are defined as in the proof of Theorem 1. By the argument used previously to show that the set \( S \) is convex, we can easily establish that \( S_2^* \) is convex. Since \( S_1 \) is convex and equations (1) - (6) in Theorem 1 have no solution for \( u \in U \), \( S_1 \) and \( S_2 \) are disjoint hence separated by a hyperplane. But \( S_2 \subset \text{closure } S_2^* \) (let \( \hat{\theta}_{D+1} \rightarrow 1 \)) so that \( S_1 \) and \( S_2 \) are also separated.
References


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