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On smooth optimal control determination

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2004-02-20

Abstract

When using the Pontryagin Maximum Principle in optimal control problems, the most difficult part of the numerical solution is associated with the non-linear operation of the maximization of the Hamiltonian over the control variables. For a class of problems, the optimal control vector is a vector function with continuous time derivatives. A method is presented to find this smooth control without the maximization of the the Hamiltonian. Three illustrative examples are considered.

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The classical optimal control problem

- Consider the classical optimal control problem (OCP), Pontryagin *et al.* (1962), Lee and Marcus (1967), Athans and Falb (1966), etc.

$$\int_0^T f_0(x, u) dt \rightarrow \min \quad (1)$$

$$\frac{dx}{dt} = f(x, u), \quad (2)$$

$$x(0) = x_0, \quad x(T) = x_T. \quad (3)$$

- the control variables $u(t) \in \mathbf{R}^m$, the state variables $x(t) \in \mathbf{R}^n$, and $f(x, u) \in \mathbf{R}^n$ are column vectors, with $m \leq n$.

- $f_0(x, u), f(x, u)$ are **smooth** in all arguments.

- The *Hamiltonian* is

$$H = p^T f(x, u) - f_0(x, u). \quad (5)$$

where it holds for the column vector $p(t) \in \mathbf{R}^n$ of co-state variables, that

$$\frac{dp}{dt} = -\frac{\partial H^T}{\partial x} = -\frac{\partial f^T}{\partial x} p + \frac{\partial f_0^T}{\partial x} \quad (6)$$

according to the Pontryagin Maximum Principle (PMP).



Classical solution

$$\int_0^T f_0(x, u) dt \rightarrow \min \quad (1)$$

$$\frac{dx}{dt} = f(x, u), \quad (2)$$

$$x(0) = x_0, \quad x(T) = x_T. \quad (3)$$

$$H = p^T f(x, u) - f_0(x, u). \quad (5)$$

$$\frac{dp}{dt} = -\frac{\partial H^T}{\partial x} = -\frac{\partial f^T}{\partial x} p + \frac{\partial f_0^T}{\partial x} \quad (6)$$

- If an optimal solution (x^*, u^*) exists, then, by PMP, it holds that $H(x^*, u^*, p^*) \geq H(x^*, u, p^*)$ implying here by smoothness, and the presence of constraint (3) only, that for $u = u^*$,

$$\frac{\partial H}{\partial u} = 0. \quad (7)$$

or, with (5) inserted into (7),

$$p^T \frac{\partial f}{\partial u} - \frac{\partial f_0}{\partial u} = 0 \quad (8)$$

where $\partial f_0 / \partial u$ is $1 \times m$, and $\partial f / \partial u$ is $n \times m$.

- To find (x^*, u^*, p^*) the two point boundary value problem (2)-(6) must be solved.
- At each t , (8) gives u^* as a function of x and p . (8) is often non-linear, and computationally costly.**
- $p(0)$ has as many unknowns as given end conditions $x(T)$.



The new idea without optimization w.r.t. u

$$\int_0^T f_0(x, u) dt \rightarrow \min \quad (1)$$

$$\frac{dx}{dt} = f(x, u), \quad (2)$$

$$x(0) = x_0, \quad x(T) = x_T. \quad (3)$$

$$H = p^T f(x, u) - f_0(x, u). \quad (5)$$

$$\frac{dp}{dt} = -\frac{\partial H^T}{\partial x} = -\frac{\partial f^T}{\partial x} p + \frac{\partial f_0^T}{\partial x} \quad (6)$$

$$\frac{\partial H}{\partial u} = 0. \quad (7)$$

$$p^T \frac{\partial f}{\partial u} - \frac{\partial f_0}{\partial u} = 0 \quad (8)$$

- We note that (8) is linear in p .
- Assume that $\text{rank}(\partial f/\partial u) = m \Rightarrow \exists$ a non-singular $m \times m$ submatrix. Then, re-index the corresponding vectors
 $x = [x^a; x^b]; p = [p^a; p^b]; f(x, u) = [f^a(x, u); f^b(x, u)]$

where \square^a denotes an m -vector. Then, (8) gives

$$p^{aT} \frac{\partial f^a}{\partial u} + p^{bT} \frac{\partial f^b}{\partial u} - \frac{\partial f_0}{\partial u} = 0 \quad (9)$$

$$p^a = -\left(\frac{\partial f^a}{\partial u}\right)^{-1} \frac{\partial f^b}{\partial u} p^b + \left(\frac{\partial f^a}{\partial u}\right)^{-1} \frac{\partial f_0}{\partial u} \doteq A(x, p^b, u) \quad (10)$$

- Hence by linear operations, m elements of $p \in \mathbf{R}^n$, i.e. p^a , are computed as a function of u, x , and p^b .



The new idea, cont'd

$$\int_0^T f_0(x, u) dt \rightarrow \min \quad (1)$$

$$\frac{dx}{dt} = f(x, u), \quad (2)$$

$$x(0) = x_0, \quad x(T) = x_T. \quad (3)$$

$$H = p^T f(x, u) - f_0(x, u). \quad (5)$$

$$\frac{dp}{dt} = -\frac{\partial H^T}{\partial x} = -\frac{\partial f^T}{\partial x} p + \frac{\partial f_0^T}{\partial x} \quad (6)$$

$$\frac{\partial H}{\partial u} = 0. \quad (7)$$

$$p^{aT} \frac{\partial f^a}{\partial u} + p^{bT} \frac{\partial f^b}{\partial u} - \frac{\partial f_0}{\partial u} = 0 \quad (9)$$

$$p^a = \left[\frac{\partial f^a}{\partial u} \right]^{-1} \left[\frac{\partial f_0}{\partial u} - \frac{\partial f^b}{\partial u} p^b \right] \doteq A(x, p^b, u) \quad (10)$$

- Differentiate (10):

$$\frac{dp^a}{dt} = \frac{\partial A}{\partial x} f(x, u) + \frac{\partial A}{\partial u} \frac{du}{dt} + \frac{\partial A}{\partial p^b} \frac{dp^b}{dt} \quad (11)$$

$$B \doteq \frac{\partial A}{\partial u}$$

- where B is assumed non-singular. (6) gives

$$\frac{dp^a}{dt} = -\frac{\partial H^T}{\partial x^a} = -\frac{\partial f^a}{\partial x^a} p^a - \frac{\partial f^b}{\partial x^a} p^b + \frac{\partial f_0^T}{\partial x^a} \quad (12)$$

$$\frac{dp^b}{dt} = -\frac{\partial H^T}{\partial x^b} = -\frac{\partial f^a}{\partial x^b} p^a - \frac{\partial f^b}{\partial x^b} p^b + \frac{\partial f_0^T}{\partial x^b} \doteq S(x, p^b, u) \quad (13)$$

- (10) into RHS of (12, 13), noting that dp^a/dt is given by the RHS of (11) and (12), and solving for du/dt , gives

$$\frac{du}{dt} = B^{-1} \left[-\frac{\partial f^a}{\partial x^a} A - \frac{\partial f^b}{\partial x^a} p^b + \frac{\partial f_0^T}{\partial x^a} - \frac{\partial A}{\partial x} f(x, u) - \frac{\partial A}{\partial p^b} \frac{dp^b}{dt} \right] \doteq F(x, p^b, u) \quad (14)$$



Theorem

Theorem: If the optimal control problem (1)-(3), $m \leq n$, has the optimal solution x^* , u^* such that u^* is smooth and belongs to the open set U , and if the Hamiltonian is given by (5), the Jacobians $\partial f^a / \partial u$ and $B = \partial p^a / \partial u$ are non-singular, then the optimal states x^* , co-states p^{*b} , and control u^* satisfy

$$\begin{aligned} \frac{du}{dt} &= F(x, p^b, u), \\ \frac{dx}{dt} &= f(x, u), \\ \frac{dp^b}{dt} &= S(x, p^b, u). \end{aligned} \quad (15)$$

with the appropriate initial conditions $u(0)=u_0$, $p^b(0)=p^b_0$ to be found.

$$\int_0^T f_0(x, u) dt \rightarrow \min \quad (1)$$

$$\frac{dx}{dt} = f(x, u), \quad (2)$$

$$x(0) = x_0, \quad x(T) = x_T. \quad (3)$$

$$H = p^T f(x, u) - f_0(x, u). \quad (5)$$

Remark: if $m=n$, then $x^a=x$, and $p^a=p$, and (15) becomes

$$\begin{aligned} \frac{dx}{dt} &= f(x, u) \\ \frac{du}{dt} &= F(x, u) \end{aligned} \quad (15')$$

Remark: The number of equations in (15) is $2n$, just as in PMP, but without the maximization of the Hamiltonian.



Example 1: Rigid body rotation

Stopping axisymmetric rigid body rotation (Athans and Falb, 1963)

$$\frac{dx}{dt} = ay + u_1, \quad (16)$$

$$\frac{dy}{dt} = -ax + u_2$$

$$x(T) = 0, \quad y(T) = 0 \quad (17)$$

$$J = \frac{1}{4} \int_0^T (u_1^2 + u_2^2) dt \rightarrow \min$$

$$p = A = \begin{bmatrix} u_1(u_1^2 + u_2^2) \\ u_2(u_1^2 + u_2^2) \end{bmatrix} \quad \frac{dp_x}{dt} = ap_y$$

$$B = \begin{pmatrix} 3u_1^2 + u_2^2 & 2u_1 u_2 \\ 2u_1 u_2 & u_1^2 + 3u_2^2 \end{pmatrix} \quad \frac{dp_y}{dt} = -ap_x$$

$$B^{-1} = \frac{1}{(3[u_1^2 + u_2^2]^2)} \begin{pmatrix} u_1^2 + 3u_2^2 & -2u_1 u_2 \\ -2u_1 u_2 & 3u_1^2 + u_2^2 \end{pmatrix}$$

$$\frac{du_1}{dt} = au_2, \quad (23)$$

$$\frac{du_2}{dt} = -au_1$$

$$H = p^T f(x, u) - f_0(x, u). \quad (5) \quad p^a = \left[\frac{\partial f^a T}{\partial u} \right]^{-1} \left[\frac{\partial f_0^T}{\partial u} - \frac{\partial f^b T}{\partial u} p^b \right] \triangleq A(x, p^b, u) \quad (10)$$

$$\frac{dp}{dt} = -\frac{\partial H^T}{\partial x} = -\frac{\partial f^T}{\partial x} p + \frac{\partial f_0^T}{\partial x} \quad (6) \quad B \triangleq \frac{\partial A}{\partial u} \quad \begin{aligned} \frac{dx}{dt} &= f(x, u) \\ \frac{du}{dt} &= F(x, u) \end{aligned} \quad (15')$$

$$\frac{du}{dt} = B^{-1} \left[-\frac{\partial f^a T}{\partial x^a} A - \frac{\partial f^b T}{\partial x^a} p^b + \frac{\partial f_0^T}{\partial x^a} - \frac{\partial A}{\partial x} f(x, u) - \frac{\partial A}{\partial p^b} \frac{dp^b}{dt} \right] \triangleq F(x, p^b, u) \quad (14)$$



Example 1: Rigid body rotation, cont'd

$$\frac{dx}{dt} = ay + u_1, \quad (16)$$

$$\frac{dy}{dt} = -ax + u_2 \quad (16)$$

$$x(T) = 0, y(T) = 0 \quad (17)$$

$$\frac{du_1}{dt} = au_2, \quad (23)$$

$$\frac{du_2}{dt} = -au_1$$

• (23) \Rightarrow

$$u_1^2 + u_2^2 = C^2$$

• Polar coordinates: $\Rightarrow (u_1, u_2)$ and (x, y) rotate collinearly with the same angular velocity a .

$$x = r \sin \theta \quad (24)$$

$$y = r \cos \theta$$

• then, clearly,

$$u_1 = -C \frac{x}{\sqrt{x^2 + y^2}}$$

$$u_2 = -C \frac{y}{\sqrt{x^2 + y^2}} \quad (25)$$

$$\frac{dr}{dt} = -C$$

$$\frac{d\theta}{dt} = a. \quad (26)$$

• Let

$$\sqrt{x(0)^2 + y(0)^2} = R$$

then, from (17), (25), (26),

$$u_1(0)/C = -x(0)/R$$

$$u_2(0)/C = -y(0)/R$$

$$C = R/T$$

The problem is solved without maximizing the Hamiltonian!



Example 2: Optimal spacing for greenhouse lettuce growth

Optimal variable spacing policy (Seginer, Ioslovich, Gutman), assuming constant climate

$$\frac{dv}{dt} = \frac{v}{W} G(W), \quad (30)$$

$$J = \int_0^T \frac{v}{W} c_R dt$$

$$v(T) = v_T \quad (31)$$

$$v = aW$$

with v [kg/plant] = dry mass, G [kg/m²/s] net photosynthesis, W [kg/m²] plant density (control), a [m²/plant] spacing, v_T marketable plant mass, and final time T [s] free.

$$H = \frac{v}{W} (pG(W) - c_R) \quad (32) \quad \bullet \quad (36), (37) \Rightarrow$$

$$\frac{\partial H}{\partial W} = -\frac{v}{W^2} (pG(W) - c_R) + \frac{vp}{W} \frac{\partial G}{\partial W} \quad \frac{dW}{dt} = -\frac{\frac{\partial G}{\partial W} (G - W \frac{\partial G}{\partial W})}{W \frac{\partial^2 G}{\partial W^2}} \quad (38)$$

• p is obtained from $\partial H / \partial W = 0$, Free final time $\Rightarrow H(T) = 0$ (39)

$$p = \frac{c_R}{G(W) - W \frac{\partial G}{\partial W}} \quad (34) \quad \bullet \quad \text{Then, at } t=T, (39, 32, 34):$$

$$\frac{dp}{dt} = -\frac{\partial H}{\partial v} = -\frac{pG(W) - c_R}{W} \quad (35) \quad v \frac{c_R \frac{\partial G}{\partial W}}{(G - W \frac{\partial G}{\partial W})} = 0 \quad (40)$$

• (34), (35) \Rightarrow (38, 40) show that $\forall t$, W^* satisfies

$$\frac{dp}{dt} = -\frac{c_R \frac{\partial G}{\partial W}}{(G - W \frac{\partial G}{\partial W})} \quad (36) \quad \frac{\partial G(W^*)}{\partial W^*} = 0$$

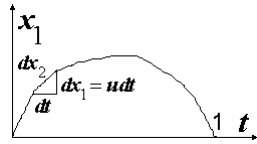
• Differentiating (34), For free final time, (38)... also w/o maximization of the Hamiltonian, I&G 99

$$\frac{dp}{dt} = \frac{dW}{dt} \frac{c_R W \frac{\partial^2 G}{\partial W^2}}{(G - W \frac{\partial G}{\partial W})^2} \quad (37)$$



Example 3: Maximal area under a curve of given length

Variational, isoparametric problem,
e.g. Gelfand and Fomin. (1969).



$$\int_0^1 x_1 dt \rightarrow \max \quad (42)$$

$$\frac{dx_1}{dt} = u, \quad (43)$$

$$\frac{dx_2}{dt} = \sqrt{1+u^2}. \quad (44)$$

$$x_1(0) = x_1(1) = 0$$

$$x_2(0) = 0, \quad x_2(1) = \pi/3 \quad (45)$$

$$H = p_1 u + p_2 \sqrt{1+u^2} + x_1 \quad (46)$$

$$\frac{dp_1}{dt} = -1 \quad (47)$$

$$\frac{dp_2}{dt} = 0.$$

$$\frac{\partial H}{\partial u} = p_1 + p_2 \frac{u}{\sqrt{1+u^2}} = 0 \quad (48)$$

- Here, it is possible to solve (48) for u , but let us solve for p_1

$$p_1 = -p_2 \frac{u}{\sqrt{1+u^2}} \quad (49)$$

- Differentiating (49), and using (47) gives

$$\frac{du}{dt} = \frac{(1+u^2)^{3/2}}{p_2} \quad (50)$$

- Guessing $p_2 = \text{constant}$ and $u(0)$, and integrate (43), (44), (50), such that (45) is satisfied, yields

$$p_2 = -1 \quad u(0) = 1/\sqrt{3} \quad !$$



Conclusions

- A method to find the smooth optimal control for a class of optimal control problems was presented.
- The method does not require the maximization of the Hamiltonian over the control.
- Instead, the ODEs for m co-states are substituted for ODEs for the m smooth control variables.
- Three illustrative examples were given.