

## Sufficient conditions for a minimum in a classical optimal control problem

by

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Based on the classical methods of the variational calculus a practical method of construction of an optimal feedback control has been given in the paper. In consequence, it yields Weierstrass sufficient conditions for a minimum.

### Introduction

The control theory literature has been concerned mainly with sufficient conditions for global optimality, the most well-known result being the Hamilton-Jacobi-Bellman theorem [2]. This, however, requires the existence of a feedback control and a solution to the Hamilton-Jacobi partial differential equation. The local theory treats mainly the exact expression  $\Delta I$  of a change in a functional due to a change in control [3]. A common technique is to add to the integrand of the functional an exact differential in order to transform the original problem into a simpler one (see, e.g. [2]; on the classical technique [1]).

In the calculus of variations there exists a technique of calculating that exact differential. It is based on the theory of the field.

The present paper examines the case when the differential form  $ydx-Hdt$  becomes exact in a control problem, which then gives a method of constructing an optimal feedback control and, in consequence, the Weierstrass sufficiency theorem.

### 1. Preliminary notes and the algorithm of Huygens for on optimal control

1. We shall be dealing with a Borel subset  $U$  of  $R^r$ . The assumption that  $U$  is Borel is connected with the maximum principle. A Lebesgue measurable function  $u: [0, 1] \rightarrow U$  will be called a control or an admissible control. The choice of the time interval  $[0, 1]$  is a normalization for notational convenience.

Let  $f(t, x, u)$  be a vector function  $f: [0, 1] \times R^n \times R^r \rightarrow R^n$ , and let  $L(t, x, u)$  be a scalar function defined on  $[0, 1] \times R^n \times R^r$ .

We suppose, for each  $t$  in  $[0, 1]$  and  $u$  in  $U$ , that the functions  $(x) \rightarrow f(t, x, u)$ ,  $(x) \rightarrow L(t, x, u)$  are  $C^1$  in  $R^n$ , and the functions  $(t, x, u) \rightarrow f(t, x, u)$ ,  $(t, x, u) \rightarrow L(t, x, u)$  are continuous in  $[0, 1] \times R^n \times R^r$ .

An admissible trajectory  $x$  corresponding to a control  $u$  is an absolutely continuous function  $x: [0, 1] \rightarrow R^n$  satisfying

$$\dot{x}(t) = f(t, x(t), u(t)), \text{ almost everywhere in } [0, 1]. \quad (1.1)$$

2. The class of all admissible pairs  $(x(t), u(t))$ ,  $t \in [0, 1]$ , such that in the first extremity  $x(0) = 0$  and in the second  $x(1) = e$  where  $e$  is a fixed point in  $R^n$ , will be denoted by  $M$ .

Our main problem is the following:

Find a minimum of the integral

$$I(x, u) = \int_0^1 L(\tau, x(\tau), u(\tau)) d\tau \quad (1.2)$$

“along” the pair  $(x(t), u(t))$  from among all pairs of  $M$ .

Further, we shall consider not only a single minimum problem, but various minimum problems which arise when the second extremity of a trajectory takes different positions in the space  $R^{n+1}$  of the variables  $(t, x)$ . The integrand  $L$  and the first extremity will be kept fixed.

To this effect, we denote by  $M(t, x)$  the class of all admissible pairs  $(x(\tau), u(\tau))$ ,  $\tau \in [0, t]$ , such that  $x(0) = 0$ ,  $x(t) = x$ ,  $0 < t \leq 1$ ,  $x \in R^n$ . The same symbol  $x$  used for a trajectory and a point of  $R^n$  will not lead to confusion, for its meaning will be clear from the context. In the minimum problem for this class we take the integral in (1.2) in the interval  $[0, t]$ , i.e.

$$I^t(x, u) = \int_0^t L(\tau, x(\tau), u(\tau)) d\tau. \quad (1.3)$$

3. A pair  $(x(\tau), u(\tau))$  of  $M(t, x)$  will be said to satisfy the maximum principle (see, e.g. [2]) if there exists an absolutely continuous vector function  $y: [0, 1] \rightarrow R^n$  such that

$$\begin{aligned} \text{(i)} \quad \dot{y}(\tau) &= -f_x(\tau, x(\tau), u(\tau)) y(\tau) + \lambda L_x(\tau, x(\tau), u(\tau)), \\ \text{(ii)} \quad y(\tau) f(\tau, x(\tau), u(\tau)) - \lambda L(\tau, x(\tau), u(\tau)) &= \\ &= \sup \{ y(\tau) f(\tau, x(\tau), u) - \lambda L(\tau, x(\tau), u) : u \in U \} \end{aligned} \quad (1.4)$$

where we consider only the situation when  $\lambda = 1$ .

4. Suppose now that, in any way (for example from the maximum principle and the existence theorems), we are able to suspect that, for the pair  $(x^0(\tau), u^0(\tau))$ ,

$\tau \in [0, t]$  of  $M(t, x)$ , integral (1.3) attains its minimum in the class  $M(t, x)$ . We denote the value of this suspected minimum by  $S(t, x)$ .

More generally, we can define  $S(t, x) = \inf_{(x, u) \in M(t, x)} I(x, u) = \inf_{M(t, x)} \int_0^t L(\tau, x(\tau), u(\tau)) d\tau$ . The function  $S(t, x)$  is then always defined in  $M(t, x)$  and called the value function. The  $S(t, x)$  may have the values  $-\infty$  and  $+\infty$ , the latter whenever the class  $M(t, x)$  is empty. We shall assume, however, in the whole paper that  $M \neq \emptyset$  and so is  $M(t, x)$  for some  $(t, x)$ . It is clear that the pair  $(x^0(\tau), u^0(\tau))$ ,  $\tau \in [0, t]$  and the value  $S(t, x)$  depend on  $(t, x) \in [0, 1] \times R^n$ . Thus  $S(t, x)$  now becomes a function of  $(t, x)$ . Evidently, it vanishes at the point  $(0, 0)$ , so that we can also write (1.3) along  $(x^0(\tau), u^0(\tau))$ ,  $\tau \in [0, t]$ , as the difference  $S(t, x) - S(0, 0)$ . We shall write  $\Phi_0 = \Phi_0(t, x, u)$  for the exact derivative  $S_t + \dot{x}S_x = S_t + fS_x$  of the function  $S(t, x)$  if it exists. Then  $S(t, x) - S(0, 0)$  is along  $(x^0(\tau), u^0(\tau))$  of  $M(t, x)$  the integral of  $\Phi_0$ . Of course, the pair  $(x^0(\tau), u^0(\tau))$ ,  $\tau \in [0, 1]$  must satisfy the inequality  $I(x, u) \geq I(x^0, u^0)$ , i.e., if  $\Phi_0$  exists, there is

$$\int_0^1 \{L(\tau, x(\tau), u(\tau)) - \Phi_0(\tau, x(\tau), u(\tau))\} d\tau \geq 0 \quad (1.5)$$

for all pairs  $(x, u)$  of  $M$ . This inequality is certainly true for all admissible triplets  $(t, x, u)$ ,

$$L(t, x, u) \geq \Phi_0(t, x, u). \quad (1.6)$$

5. We shall now reformulate the consideration of 4. in the following algorithm of Huygens (comp. [5]). Let  $M$  be a class of pairs  $(x, u)$  described in 2., and let  $(x^0, u^0)$  be a member of that class. If there exists an exact derivative  $\Phi_0$  such that  $L = \Phi_0$  along  $(x^0, u^0)$  and that  $L \geq \Phi_0$  along all pairs  $(x, u)$  of  $M$ , then  $I(x^0, u^0)$  is the minimum of  $I(x, u)$  for  $(x, u) \in M$ .

The proof of this algorithm is evident. Indeed,  $I(x, u)$  is, for  $(x, u) \in M$ , not less than the corresponding integral of  $\Phi_0$  and we can take the latter along  $(x^0, u^0)$  without altering its value which is then  $I(x^0, u^0)$ .

## 2. A method of feedback control

Denote by  $V$  the set in  $R^{n+1}$  covered by graphs of all admissible trajectories, and let  $D_{\bar{t}}$  be any domain in  $V$  containing the interval  $[0, 1]$ . Let, further,  $\bar{u}(t, x)$  denote a vector-valued function defined in  $D_{\bar{t}}$  with values in  $U$ .

DEFINITION 2.1. We call  $\bar{u}(t, x)$  an optimal feedback control if there exists an exact derivative

$$\Phi_0 = S_t + \dot{x}S_x = S_t + fS_x \quad (2.1)$$

such that, for all admissible triplets  $(t, x, u)$  (i.e.  $(t, x) \in D_{\bar{z}}$  and those  $u(t) \in U$  for which the graphs of the corresponding trajectories lie in  $D_{\bar{z}}$ );

$$L(t, x, u) \geq \Phi_0(t, x, u) \quad (2.2)$$

with equality when  $u = \bar{u}$ .

Let  $M_{D_{\bar{z}}}$  denote a subclass of  $M$  of those pairs  $(x, u)$  whose graphs of trajectories lie in  $D_{\bar{z}}$ . Then, by the algorithm of Huygens, it is evident, if an optimal feedback control  $\bar{u}(t, x)$  exists, that each pair  $(x^0(t), \bar{u}(t, x^0(t)))$  of  $M_{D_{\bar{z}}}$  affords a minimum to  $I(x, u)$  relative to pairs  $(x, u)$  of  $M_{D_{\bar{z}}}$ .

In 4. (§ 1) we started from the function  $S$  which, of course, determines  $\Phi_0$ . Conversely,  $\Phi_0$  determines  $S$  for an additive constant. Our task is to show that when  $\Phi_0$  exists it is also completely determined by  $\bar{u}$ .

To this effect, we observe, by (2.2), that the difference  $L - \Phi_0$  regarded as a function of  $u$  only attains its minimum equal to zero at  $u = \bar{u}$ . This requires that, if we assume  $\bar{y}(t, x) = S_x(t, x)$ , then

$$S_t = L - f S_x = L - f \bar{y} \quad (2.3)$$

at  $u = \bar{u}$ . If we put  $S_t$  of (2.3) in inequality (2.2), then the latter is known as the Hamilton-Jacobi-Bellman sufficient condition for global optimality (see [3]), and if we denote  $-H(t, x, \bar{u}(t, x), S_x) = L(t, x, \bar{u}(t, x)) - f(t, x, \bar{u}(t, x)) S_x(t, x)$ , then (2.3) becomes

$$S_t = -H(t, x, \bar{u}(t, x), S_x), \quad (2.4)$$

i.e. the Hamilton-Jacobi-Bellman differential equation. Simultaneously, by the above

$$H(t, x, \bar{u}(t, x), \bar{y}(t, x)) = \max_u \{ \bar{y}(t, x) f(t, x, u) - L(t, x, u) \} \quad (t, x) \in D_{\bar{z}} \quad (2.5)$$

or

$$H(t, x, \bar{u}(t, x), \bar{y}(t, x)) - H(t, x, u, \bar{y}(t, x)) \geq 0 \quad (2.6)$$

for all admissible  $(t, x, u)$ . The left-hand side of (2.6) is known as the extension of the excess function of Weierstrass (see [4]).

We may also express (2.3) by saying that for rectifiable curves  $C$  lying in a simply connected subdomain of  $D_{\bar{z}}$ , the curvintegral

$$\int_C (L(t, x, \bar{u}(t, x)) - f(t, x, \bar{u}(t, x)) \bar{y}(t, x)) dt + \bar{y}(t, x) dx \quad (2.7)$$

depends only on the endpoints of  $C$ . This expression is termed Hilbert's independent integral: its value, on account of (2.3), is clearly  $S(B) - S(A)$  where  $A, B$  are the endpoints of  $C$ .

We shall term  $D_{\bar{z}}$ -levels the set of points  $P \in D_{\bar{z}}$  on which  $S(P)$  is constant. If we denote the left-hand side of (2.6), by  $E$ , then we have  $L = \Phi_0 + E$  and therefore, for

any pair  $(x, u)$  of  $M$  whose arc of the trajectory  $C$  joins in  $D_{\bar{u}}$  two levels  $S=S_1$  and  $S=S_2$ ,

$$\int_C L(t, x, u) dt = S_2 - S_1 + \int_C E dt. \quad (2.8)$$

This is an extension of the famous formula of Weierstrass, whose equivalent in the classical calculus, according to Young [5], has revolutionized the variational calculus. In contrast, to date, formula (2.8) or its analogon, that is, the exact expression for  $\Delta I = I(x, u) - S(1, e)$  (see [4]), do not appear to have played a central role in modern control theory, although the expressions  $\Delta I$  do appear as tools in the derivation of optimization algorithms (see [3]).

### 3. The necessary exactness condition

The exactness condition that we have to study is that the expression

$$\bar{y}(t, x) dx + (L(t, x, u) - \bar{y}(t, x) \dot{x}) dt \quad (3.1)$$

reduces to an exact differential  $dS$  in the variables  $(t, x)$  when we substitute  $\dot{x} = f(t, x, u)$  and  $u = \bar{u}(t, x)$ ;  $\bar{y}$  is defined as in § 2. We wish to find out when this occurs.

To this effect, we shall enlarge our setting by using the transformation

$$(t, \sigma) \rightarrow (t, x), \quad (t, x) \in D_{\bar{u}}, \quad (3.2)$$

in which the time  $t$  is unchanged. This means that  $x$  is replaced by a function  $x(t, \sigma)$  where  $(t, \sigma)$  belongs to some open simply connected set  $(0, 1) \times G$  of  $R^{m+1}$ . Next, we enlarge (3.2) to a map of triplets

$$(t, \sigma, \dot{\sigma}) \rightarrow (t, x, \dot{x}) \quad (3.2')$$

by putting  $\dot{x} = x_t + x_\sigma \dot{\sigma}$ , i.e. we substitute for the variable  $\dot{x}$  the total derivative of  $x(t, \sigma)$ . Evidently, now

$$f(t, x(t, \sigma), u) = x_t + x_\sigma \dot{\sigma}. \quad (3.3)$$

Further, the dot over a function of  $(t, \sigma)$  will always denote the total derivative. We suppose that the function  $x(t, \sigma)$  defined in (3.2') is  $C^2$  in  $(0, 1) \times G$ , and  $\bar{y}(t, \sigma) = \bar{y}(t, x(t, \sigma))$  is  $C^1$  there.

Now, we study under what additional circumstances a family of curves of  $(0, 1) \times G$  given by a differential equation

$$x_\sigma(t, \sigma) \dot{\sigma} = -x_t(t, \sigma) + f(t, x(t, \sigma), \bar{u}(t, x(t, \sigma))) \quad (3.3')$$

is such that map (3.2') turns expression (3.1) into an exact differential  $dS$  in the variables  $(t, \sigma)$ . Then, the integral of (3.1) on an arbitrary curve of  $(0, 1) \times G$  depends only on the endpoints. We call it the corresponding Hilbert independent integral.

We shall suppose that, for the functions  $\tilde{L}(t, \sigma) = L(t, x(t, \sigma), \bar{u}(t, x(t, \sigma)))$ ,  $\tilde{f}(t, \sigma) = f(t, x(t, \sigma), \bar{u}(t, x(t, \sigma)))$ , there exist continuous derivatives  $\tilde{L}_\sigma(t, \sigma)$ ,  $\tilde{f}_\sigma(t, \sigma)$  in  $(0, 1) \times G$  and  $\frac{\partial}{\partial \sigma} L(t, x, \bar{u}(t, x(t, \sigma)))$ ,  $\frac{\partial}{\partial \sigma} f(t, x, \bar{u}(t, x(t, \sigma)))$ , for each fixed  $(t, x) \in D_{\bar{x}}$ , in  $(0, 1) \times G$  and they satisfy at  $(t, x)$ ,  $x = x(t, \sigma)$ , the relations:

$$\begin{aligned} \frac{\partial \tilde{L}}{\partial \sigma} &= \frac{\partial}{\partial \sigma} L(t, x, \bar{u}(t, x(t, \sigma))) + L_x(t, x, \bar{u}(t, x(t, \sigma))) x_\sigma(t, \sigma), \\ \frac{\partial \tilde{f}}{\partial \sigma} &= \frac{\partial}{\partial \sigma} f(t, x, \bar{u}(t, x(t, \sigma))) + f_x(t, x, \bar{u}(t, x(t, \sigma))) x_\sigma(t, \sigma). \end{aligned} \quad (3.3'')$$

We note at once that the induced feedback control  $\bar{u}(t, x(t, \sigma))$  expressed by map (3.2) satisfies, for  $(t, \sigma) \in (0, 1) \times G$ , the induced maximum relation

$$\begin{aligned} \bar{y}(t, x(t, \sigma)) f(t, x(t, \sigma), \bar{u}(t, x(t, \sigma))) - L(t, x(t, \sigma), \bar{u}(t, x(t, \sigma))) = \\ = \max_u \{ \bar{y}(t, x(t, \sigma)) f(t, x(t, \sigma), u) - L(t, x(t, \sigma), u) \}. \end{aligned} \quad (3.3''')$$

It is clear that one of the necessary and sufficient conditions for the existence of  $S(t, \sigma)$  is satisfaction of the relations

$$S_{\sigma_i \sigma_j} = S_{\sigma_j \sigma_i}, \quad S_{\sigma_i t} = S_{t \sigma_i}, \quad i, j = 1, \dots, m, \quad \sigma = (\sigma_1, \dots, \sigma_m). \quad (3.4)$$

Let  $1 \leq i, j \leq m$  be arbitrarily fixed. After substitution of (3.2') in (3.1) we easily get from the expression obtained that

$$S_{\sigma_i \sigma_j} - S_{\sigma_j \sigma_i} = \frac{\partial}{\partial \sigma_j} (\bar{y} x_{\sigma_i}) - \frac{\partial}{\partial \sigma_i} (\bar{y} x_{\sigma_j}) = x_{\sigma_i} \bar{y}_{\sigma_j} - x_{\sigma_j} \bar{y}_{\sigma_i}, \quad (3.5)$$

$$S_{\sigma_i t} - S_{t \sigma_i} = \frac{\partial}{\partial t} (\bar{y} x_{\sigma_i}) - \frac{\partial}{\partial \sigma_i} (\bar{y} x_t + (\tilde{L} - \bar{y} \dot{x})). \quad (3.6)$$

We calculate exactly the right-hand sides of (3.5) and (3.6). Thus

$$\frac{\partial}{\partial t} (\bar{y} x_{\sigma_i}) = \bar{y} x_{\sigma_i t} + x_{\sigma_i} \dot{\bar{y}} = x_{\sigma_i} \dot{\bar{y}} + \bar{y} x_{\sigma_i t} + x_{\sigma_i} (\bar{y}_t - \dot{\bar{y}}), \quad (3.7)$$

$$\frac{\partial}{\partial \sigma_i} (\bar{y} x_t) = \bar{y}_{\sigma_i} x_t + \bar{y} x_{\sigma_i t} = \bar{y}_{\sigma_i} \dot{x} + \bar{y} x_{\sigma_i t} + \bar{y}_{\sigma_i} (x_t - \dot{x}). \quad (3.8)$$

Note that, by (3.5),

$$\begin{aligned} x_{\sigma_i} (\dot{\bar{y}} - \bar{y}_t) - \bar{y}_{\sigma_i} (x_t - \dot{x}) &= - \sum_{j=1}^m (x_{\sigma_i} \bar{y}_{\sigma_j} - \bar{y}_{\sigma_i} x_{\sigma_j}) \dot{\sigma}_j = \\ &= - \sum_{j=1}^m (S_{\sigma_i \sigma_j} - S_{\sigma_j \sigma_i}) \dot{\sigma}_j. \end{aligned} \quad (3.9)$$

We also note that, since (2.5) is satisfied (comp. also (3.3''')), remembering (3.3),

$$\begin{aligned} \frac{\partial}{\partial \sigma_i} (\tilde{y}\dot{x} - \tilde{L}) &= \frac{\partial}{\partial \sigma_i} (\tilde{y}\tilde{f} - \tilde{L}) = \tilde{y}_{\sigma_i} \dot{x} + \tilde{y}f_x x_{\sigma_i} - L_x x_{\sigma_i} + \frac{\partial}{\partial \sigma_i} \{ \tilde{y}(t, x) \times \\ &\times f(t, x, \tilde{u}(t, x(t, \sigma))) - L(t, x, \tilde{u}(t, x(t, \sigma))) \} = \tilde{y}_{\sigma_i} \dot{x} + \tilde{y}f_x x_{\sigma_i} - L_x x_{\sigma_i} \end{aligned} \quad (3.10)$$

where the  $t, x$  in brackets are momentarily fixed points. By substituting (3.7), (3.8) and (3.10) in (3.6), we find that

$$S_{\sigma_i t} - S_{t \sigma_i} = x_{\sigma_i} [\tilde{y} + \tilde{y}f_x - L_x] - \sum_{j=1}^m (S_{\sigma_i \sigma_j} - S_{\sigma_j \sigma_i}) \dot{\sigma}_j.$$

Hence

$$x_{\sigma_i} [\tilde{y} + \tilde{y}f_x - L_x] = (S_{\sigma_i t} - S_{t \sigma_i}) + \sum_{j=1}^m (S_{\sigma_i \sigma_j} - S_{\sigma_j \sigma_i}) \dot{\sigma}_j. \quad (3.11)$$

Here, by (3.4), the summand on the right vanishes and, therefore, so do both sides of (3.11) for each  $i, i=1, \dots, m$ . If we change the order of the two factors on the left, the resulting set of equations, obtained from the vanishing of the left-hand side for various  $i$ , may now be written as the single vector equation

$$[\tilde{y} + \tilde{y}f_x - L_x] x_{\sigma} = 0. \quad (3.12)$$

This equation will be termed the induced conjugate differential equation.

The induced triplet  $(\sigma, u^*, y^*)$ , when  $\sigma, u^* = \tilde{u}(t, x(t, \sigma)), y^* = y^*(t, \sigma)$  satisfy (3.3'), (3.3'''), (3.12), respectively, will be referred to as satisfying the induced maximum principle.

We have thus proved the following

**THEOREM 3.1.** *Let transformation (3.2) be such that  $x(t, \sigma)$  is  $C^2$  in  $(0, 1) \times G$ ,  $\tilde{y}(t, \sigma)$  is  $C^1$  there, and there exist  $\tilde{L}_{\sigma}, \tilde{f}_{\sigma}$  that satisfy relations (3.3'''). Moreover, (3.2) turns the expression (3.1) into an exact differential  $dS$  in  $(t, \sigma)$ . Then, for each induced trajectory  $\sigma$  under control  $\tilde{u}(t, \sigma) = \tilde{u}(t, x(t, \sigma))$ , i.e., satisfying (3.3'), there exists an induced conjugate function  $y(t, \sigma)$  such that the triplet  $(\sigma, \tilde{u}, y)$  satisfies the induced maximum principle. The triplets  $(\sigma, \tilde{u}(t, \sigma), \tilde{y}(t, \sigma))$  also satisfy the induced maximum principle.*

Let now (3.2) be the identity map, so that  $x_{\sigma}$  is the identity matrix and let the assumptions of theorem 3.1 be satisfied; setting there the variable  $x$  instead of  $\sigma$  and taking the open simply connected set  $D_{\bar{x}}$  instead of  $(0, 1) \times G$ , we easily conclude that:

**THEOREM 3.1'.** *If  $x^0(t), t \in (0, 1)$ , satisfies*

$$\dot{x}^0(t) = f(t, x^0(t), \tilde{u}(t, x^0(t))),$$

then the function  $y^0(t) = \bar{y}(t, x^0(t)) = S_x(t, x^0(t))$ ,  $t \in (0, 1)$ , satisfies the conjugate vector differential equation

$$\dot{y}^0(t) = -y^0(t) f_x(t, x^0(t), \bar{u}(t, x^0(t))) + L_x(t, x^0(t), \bar{u}(t, x^0(t))).$$

#### 4. The sufficient exactness condition

In this section we give conditions for the following functions defined in  $(0, 1) \times G$  to vanish, where we assume  $m=n$

$$\varphi_i(t, \sigma) = S_{\sigma_i t}(t, \sigma) - S_{t \sigma_i}(t, \sigma) \quad i=1, \dots, n, \quad (4.1)$$

$$\psi_{ij}(t, \sigma) = S_{\sigma_i \sigma_j}(t, \sigma) - S_{\sigma_j \sigma_i}(t, \sigma), \quad i, j=1, \dots, n. \quad (4.1')$$

We put  $\psi_i = (\psi_{i1}, \dots, \psi_{in})$ ,  $\psi_j = (\psi_{1j}, \dots, \psi_{nj})$ ,  $i, j=1, \dots, n$ .

$$\left\{ \begin{array}{l} \text{From now on, we can and shall suppose, besides the assumptions} \\ \text{of § 3, that: the family of induced triplets } N \text{ satisfies the induced} \\ \text{maximum principle, } \det(x_{\sigma}) \neq 0 \text{ and } \bar{y} \text{ is } C^2 \text{ in } (0, 1) \times G, \text{ there exist} \\ \bar{f}_{\sigma\sigma} \text{ and } \bar{L}_{\sigma\sigma}. \end{array} \right. \quad (4.2)$$

In accordance with Lagrange we shall apply the following, easy to verify identities (see, e.g. [5], § 15)

$$\frac{\partial}{\partial \sigma} \psi_{ij} - \frac{\partial}{\partial \sigma_j} \psi_i - \frac{\partial}{\partial \sigma_i} \psi_j = 0. \quad (4.3)$$

$$\frac{\partial}{\partial t} \psi_{ij} - \frac{\partial}{\partial \sigma_j} \varphi_i + \frac{\partial}{\partial \sigma_i} \varphi_j = 0. \quad (4.4)$$

Thus, because of (3.11) and (4.1), (4.1'), we have

$$\varphi_i + \dot{\sigma} \psi_i = 0. \quad (4.5)$$

The total derivatives of  $\psi_{ij}(t, \sigma)$  have the form

$$\dot{\psi}_{ij} = \frac{\partial}{\partial t} \psi_{ij} + \dot{\sigma} \frac{\partial}{\partial \sigma} \psi_{ij}. \quad (4.6)$$

Applying successively (4.3), (4.4) and (4.5) to the right-hand side of (4.6), we find, remembering (3.3'), that

$$\dot{\psi}_{ij} = -\psi_i \frac{\partial \dot{\sigma}}{\partial \sigma_j} - \psi_j \frac{\partial \dot{\sigma}}{\partial \sigma_i}, \quad i, j=1, \dots, n. \quad (4.7)$$

This is rather a classical result first obtained by Lagrange and we can encounter similar results in, for example, [5] or [1].

Note that (4.7) is a system of homogeneous linear differential equations for  $\psi_{ij}$ . Moreover, the  $\varphi_i$  are, by (4.5), homogeneous linear combinations of the  $\psi_{ij}$ . Thus we have  $\mathfrak{B}$ .

**THEOREM 4.1.** *The functions  $\varphi_i, \psi_{ij}, i, j=1, \dots, n$ , will vanish identically in  $(0, 1) \times G$  if  $\psi_{ij}$  vanish at some point of each trajectory of the family  $N$  (i.e. the curve satisfying (3.3')).*

**Proof.** It is a direct consequence of the property of solutions of a system of homogeneous linear differential equations.

In the particular case of  $\dot{\sigma}=0$ , i.e. if  $\sigma=\text{const}$  along each curve satisfying (3.3'), then along each such curve the functions  $\psi_{ij}=x_{\sigma_i} \tilde{y}_{\sigma_j} - x_{\sigma_j} \tilde{y}_{\sigma_i}$  are constant.

Now, we give sufficient condition for the vanishing of the  $\varphi_i, \psi_{ij}$ .

To this effect, let  $t=t(\sigma), \sigma \in G$ , denote some  $C^1$ -locus which cuts each trajectory  $\sigma$  of  $N$  but is not tangent to it (i.e.  $t_\sigma \dot{\sigma} \neq 1$ ) and let

$$Y(t, \sigma) d\sigma - K(t, \sigma) dt \quad (4.8)$$

where  $Y=(Y_1, \dots, Y_n)=\tilde{y}x_\sigma, -K=\tilde{y}x_t+(L-\tilde{y}\dot{x})$ , denote expression (3.1) in map (3.2').

**THEOREM 4.2.** *The functions  $\varphi_i, \psi_{ij}, i, j=1, \dots, n$ , will vanish in  $(0, 1) \times G$  if expression (4.8) is an exact differential in the variable  $\sigma$  on the locus  $t=t(\sigma), \sigma \in G$ .*

**Proof.** We first observe that, by the definition of (4.1'),  $\psi_{ii}=0, \psi_{ij}=-\psi_{ji}, i, j=1, \dots, n$ . Hence  $\sum_{i,j} \dot{\sigma}_i \psi_{ij} \dot{\sigma}_j = 0$  and, by (4.5), we have the identity

$$\varphi \dot{\sigma} = 0. \quad (4.9)$$

where  $\varphi=(\varphi_1, \dots, \varphi_n)$ .

Next, we set  $\zeta=t_\sigma \dot{\sigma}, \bar{Y}(\sigma)=(\bar{Y}_1(\sigma), \dots, \bar{Y}_n(\sigma))=Y(t(\sigma), \sigma), \bar{K}(\sigma)=K(t(\sigma), \sigma)$ . At the locus  $t=t(\sigma), \sigma \in G$ , (4.8) takes the form

$$(\bar{Y} - \bar{K}t_\sigma) d\sigma = \sum_{i=1}^n (\bar{Y}_i - \bar{K}t_{\sigma_i}) d\sigma_i. \quad (4.10)$$

Since (4.10) is exact by hypothesis, we must have

$$\frac{\partial}{\partial \sigma_j} (\bar{Y}_i - t_{\sigma_i} \bar{K}) - \frac{\partial}{\partial \sigma_i} (\bar{Y}_j - t_{\sigma_j} \bar{K}) = 0, \quad i, j=1, \dots, n. \quad (4.11)$$

Here the left-hand side is the value of the expression

$$\left( \frac{\partial}{\partial \sigma_j} + t_{\sigma_j} \frac{\partial}{\partial t} \right) (Y_i - t_{\sigma_i} K) - \left( \frac{\partial}{\partial \sigma_i} + t_{\sigma_i} \frac{\partial}{\partial t} \right) (Y_j - t_{\sigma_j} K)$$

for  $t=t(\sigma)$ , which is equivalent to

$$\left( \frac{\partial Y_i}{\partial \sigma_j} - \frac{\partial Y_j}{\partial \sigma_i} \right) + t_{\sigma_i} \left( \frac{\partial Y_i}{\partial t} + \frac{\partial K}{\partial \sigma_i} \right) - t_{\sigma_i} \left( \frac{\partial Y_i}{\partial t} + \frac{\partial K}{\partial \sigma_j} \right) \quad (4.11')$$

and may be written in terms of (4.1), (4.1') (comp. (3.5), (3.6)) in the form  $\psi_{ij} + t_{\sigma_j} \varphi_i - t_{\sigma_i} \varphi_j$ . By (4.11), this vanishes for  $t=t(\sigma)$ ,  $i, j=1, \dots, n$ . Hence we obtain the vector equation

$$\psi_i + t_{\sigma} \varphi_i = t_{\sigma_i} \varphi_i. \quad (4.12)$$

By multiplying scalarwise by  $\dot{\sigma}$  and using (4.5), the definition of  $\zeta$  and identity (4.9), we find that

$$(\zeta - 1) \varphi_i = 0, \quad i=1, \dots, n,$$

whence  $\varphi_i=0$ ,  $i=1, \dots, n$ , and thus  $\varphi=0$ , so that (4.12) reduces to  $\psi_i=0$ . This shows that, for  $t=t(\sigma)$ , all  $\psi_{ij}$  vanish and so, the assumptions of theorem 4.1 are satisfied. This completes the proof.  $\blacksquare$

NOTE 1. In the particular case of  $\dot{\sigma}=0$ , the nontangency condition is always satisfied.

NOTE 2. If the functions  $\varphi_i, \psi_{ij}$ ,  $i, j=1, \dots, n$ , vanish at the locus  $t=t(\sigma)$ ,  $\sigma \in G$ , then expression (4.8) is an exact differential in  $\sigma$  there. This is a direct consequence of the relations

$$\frac{\partial}{\partial \sigma_j} (\bar{Y}_i - t_{\sigma_i} \bar{K}) - \frac{\partial}{\partial \sigma_i} (\bar{Y}_j - t_{\sigma_j} \bar{K}) = \psi_{ij} + t_{\sigma_j} \varphi_i - t_{\sigma_i} \varphi_j.$$

COROLLARY 4.1. *If the assumptions of theorem 4.2 are satisfied, the function  $S(\bar{t}, \bar{\sigma})$ ,  $(\bar{t}, \bar{\sigma}) \in (0, 1) \times G$ , exists and may be defined by*

$$S(\bar{t}, \bar{\sigma}) = - \int_{\bar{t}}^{t(\sigma)} L(t, x(t, \sigma), \bar{u}(t, \sigma)) dt$$

where the integral is taken along the trajectory  $\sigma$  of  $N$  passing through the point  $(\bar{t}, \bar{\sigma})$  and cutting the locus  $t=t(\sigma)$ ,  $\sigma \in G$ .

This is a direct consequence of the integration of (4.8) along  $\sigma$  in the interval  $[\bar{t}, t(\sigma)]$ .

## 5. The practical method of constructing an optimal feedback control

In the preceding sections we gave necessary and sufficient conditions for the independence of the Hilbert integral

$$\int \bar{y}(t, x) dx + (L(t, x, \bar{u}(t, x)) - \bar{y}(t, x) f(t, x, \bar{u}(t, x))) dt \quad (5.1)$$

of the  $(t, \sigma)$  path of integration for given ends. They are as follows: each induced triplet  $(\sigma, u^*, y^*)$  satisfies the induced maximum principle and at some point of

this triplet the functions  $\psi_{ij}$ ,  $\varphi_i$  vanish. All that was true by the assumption that there exist a feedback control  $\bar{u}(t, x)$ ,  $(t, x) \in D_{\bar{u}}$  and a function  $\bar{y}(t, x)$ ,  $(t, x) \in D_{\bar{y}}$ .

In this section we give the method of a practical construction of both functions  $\bar{u}$  and  $\bar{y}$ .

To this effect, we shall describe a special family of induced triplets. Thus, let us define on an open simply connected set  $G \subset R^n$  a pair of  $C^1$ -functions  $t^-(\sigma)$ ,  $t^+(\sigma)$ ,  $t^-(\sigma) < t^+(\sigma)$ , with values in the interval  $[0, 1]$  and such that, for some  $\sigma^0 \in G$ ,  $t^-(\sigma^0) = 0$ ,  $t^+(\sigma^0) = 1$ .

Denote by  $Z^-$ ,  $Z$ ,  $Z^+$  the sets of pairs  $(t, \sigma)$  where  $\sigma \in G$  and  $t$  satisfies, respectively, the conditions

$$0 \leq t^-(\sigma) = t, \quad t^-(\sigma) < t < t^+(\sigma), \quad t = t^+(\sigma) \leq 1. \quad (5.2)$$

The notation  $[Z]$  will be used for the union of the sets  $Z^-$ ,  $Z$ ,  $Z^+$ .

Now, we consider a family  $\Sigma$  of pairs  $(x, u)$  depending on a parameter  $\sigma$ , which satisfy the maximum principle (comp. 3. § 1), given by the functions

$$x(t, \sigma), \quad u(t, \sigma) \quad (t, \sigma) \in Z. \quad (5.3)$$

Here  $\sigma$  is the parameter which distinguishes a member of the family, i.e.  $\sigma$  remains constant on each member of  $\Sigma$ , and this member is then defined in the open interval  $t^-(\sigma) < t < t^+(\sigma)$ .

Further, we consider the set  $\tilde{G} \subset R^{2n}$  of  $(\sigma, \rho)$  and suppose that the set  $G$  is a projection of the set  $\tilde{G}$  in the following sense:

$$\left\{ \begin{array}{l} \text{Given any point } (\sigma^1, \rho^1) \in \tilde{G} \text{ and any sufficiently small open neigh-} \\ \text{bourhood } Q \subset G \text{ of } \sigma^1, \text{ there exists in } Q \text{ a } C^2\text{-function } \rho(\sigma) \text{ such} \\ \text{that } \rho(\sigma^1) = \rho^1 \text{ and that all points of the form } [\sigma, \rho(\sigma)] \text{ for } \sigma \in Q \\ \text{lie in } \tilde{G}. \end{array} \right. \quad (5.4)$$

Similarly as above, we denote by  $Z^{*-}$ ,  $Z^*$ ,  $Z^{*+}$  the sets of  $(t, \sigma, \rho)$  for which  $t$  is subject to respective conditions (5.2), and  $(\sigma, \rho) \in \tilde{G}$ . We write  $[Z^*]$  for the union of the three sets. We shall further denote by  $\Sigma^*$  a family of canonical triplets  $(x, u, y)$  which correspond to a member of  $\Sigma$  and which are obtained by giving, with functions (5.3), a further conjugate vector function

$$y(t, \sigma, \rho) \quad (t, \sigma, \rho) \in Z^*. \quad (5.5)$$

The parameter  $\rho$ , occurring in (5.5), distinguishes the corresponding canonical triplets. It appears here since, for a conjugate function which satisfies differential equation (1.4) (i), we have no additional boundary conditions.

The definitions of the functions  $x(t, \sigma)$ ,  $y(t, \sigma, \rho)$  will be supposed extended to the sets  $[Z]$ ,  $[Z^*]$ . This means defining them for  $t = t^+(\sigma)$  and  $t = t^-(\sigma)$ , where the values of  $x, y$  correspond to the end points of our members. The sets of pairs  $(t, x)$ , where  $x = x(t, \sigma)$  with  $(t, \sigma)$  belonging to  $Z^-$ ,  $Z$ ,  $Z^+$ ,  $[Z]$ , will be denoted by  $D^-$ ,  $D$ ,  $D^+$ ,  $[D]$ , respectively.

Moreover, we suppose the following conditions satisfied:

- (i) For the functions  $\tilde{L}(t, \sigma) = L(t, x(t, \sigma), u(t, \sigma))$ ,  $\tilde{f}(t, \sigma) = f(t, x(t, \sigma), u(t, \sigma))$ , there exist continuous derivatives  $\tilde{L}_{\sigma\sigma}$ ,  $\tilde{L}_{\sigma x}$ ,  $\tilde{f}_{\sigma}$ ,  $\tilde{f}_{\sigma x}$  in  $[Z]$  and  $\frac{\partial}{\partial \sigma} L(t, x, u(t, \sigma))$ ,  $\frac{\partial}{\partial \sigma} f(t, x, u(t, \sigma))$  for each fixed  $(t, x) \in D$ , in  $Z$  and they satisfy at  $(t, x)$ ,  $x = x(t, \sigma)$ , the relations:
- $$\frac{\partial \tilde{L}}{\partial \sigma} = \frac{\partial}{\partial \sigma} L(t, x, u(t, \sigma)) + L_x(t, x, u(t, \sigma)) \times x_{\sigma}(t, \sigma), \quad (5.6)$$
- $$\frac{\partial \tilde{f}}{\partial \sigma} = \frac{\partial}{\partial \sigma} f(t, x, u(t, \sigma)) + f_x(t, x, u(t, \sigma)) x_{\sigma}(t, \sigma).$$
- (ii) The function  $\tilde{y}(t, \sigma) = y(t, \sigma, \rho(\sigma))$  is  $C^2$  in  $[Z]$ .
- (iii) The function  $x(t, \sigma)$  is  $C^2$  in  $[Z]$ .
- (iv)  $\det(x_{\sigma}) \neq 0$  in  $Z^- \cup Z$  and through each point of  $D^- \cup D$  there passes one and only one trajectory  $x$  of  $\Sigma$ .

By assumptions (5.6) (iii) and (iv), the mapping

$$(t, \sigma) \rightarrow (t, x(t, \sigma)): Z^- \cup Z \rightarrow D^- \cup D \quad (5.7)$$

is a  $C^2$ -diff of  $Z^- \cup Z$  onto  $D^- \cup D$  with the inverse  $C^2$ -diff

$$\theta: (t, x) \rightarrow (t, \sigma(t, x)): D^- \cup D \rightarrow Z^- \cup Z. \quad (5.8)$$

For  $(t, x) \in D^- \cup D$ , let us set

$$\tilde{u}(t, x) = u(t, \sigma(t, x)), \quad (5.9)$$

$$\tilde{y}(t, x) = y(t, \sigma(t, x), \rho(\sigma(t, x))) = \tilde{y}(t, \sigma(t, x)). \quad (5.10)$$

Of course, (5.10) is defined only locally. We extended  $\tilde{u}(t, x)$  and  $\tilde{y}(t, x)$  to the set  $D^+$  taking there, for  $\tilde{u}(t, x)$  and  $\tilde{y}(t, x)$ , any value of  $u(t^+(\sigma), \sigma)$  and  $y(t^+(\sigma), \sigma, \rho(\sigma))$ , respectively, for that  $\sigma$  for which  $t = t^+(\sigma)$ ,  $x = x(t^+(\sigma), \sigma)$ . It is clear that  $\tilde{u}(t, x)$ ,  $\tilde{y}(t, x)$  satisfy (2.5) in  $[D]$ .

Now, we note that map (5.7) may be considered as transformation (3.2) and that the  $x(t, \sigma)$ ,  $\tilde{y}(t, x(t, \sigma)) = y(t, \sigma, \rho(\sigma)) = \tilde{y}(t, \sigma)$  defined here satisfy the assumptions about them made in § 4, i.e. (4.2). Moreover, in (3.2') we now set  $\tilde{\sigma} = 0$ , so that (3.3') takes the form

$$x_t(t, \sigma) = f(t, x(t, \sigma), u(t, \sigma)), \quad (5.11)$$

and (3.12), since  $\det(x_{\sigma}) \neq 0$ , takes the form

$$y_t(t, \sigma, \rho(\sigma)) = -y(t, \sigma, \rho(\sigma)) f_x(t, x(t, \sigma), u(t, \sigma)) + L_x(t, x(t, \sigma), u(t, \sigma)). \quad (5.12)$$

Hence we conclude that the induced triplets  $(\sigma, u^*, y^*)$ , where  $\sigma = \text{const}$ ,  $u^* = \tilde{u}(t, x(t, \sigma)) = u(t, \sigma)$ ,  $y^* = \tilde{y}(t, x(t, \sigma)) = y(t, \sigma, \rho(\sigma))$ , satisfy the induced maximum principle.

Suppose further that the Hilbert differential  $\bar{y}dx + (L - \bar{y}f) dt$  is an exact differential in the variable  $\sigma$  on the locus  $t=t^+(\sigma)$ ,  $\sigma \in G$ , i.e. in  $Z^+$ . By theorem 4.2 and note 1 following it, the functions  $\psi_{t,t}$ ,  $\varphi_t$  vanish in  $[Z]$ . Thus, Hilbert's integral (5.1) is dependent of the  $(t, \sigma)$  path of integration lying in  $[Z]$  for given ends (see note 2 § 4).

Families of pairs  $(x, u)$  satisfying the maximum principle (having form (5.3) they satisfy conditions (5.6)), for which Hilbert's integral (5.1) is independent of the  $(t, x)$  path joining in  $D^- \cup D$  two points of  $D^- \cup D$ , will be called sprays of extremal pairs, and the corresponding families of canonical triplets  $(x, u, y)$  — canonical sprays of extremal triplets. We state an existence theorem for canonical sprays.

**THEOREM 5.1.** *If the family  $\Sigma^*$ , described above, satisfies conditions (5.6), then Hilbert's integral (5.1) with  $\bar{u}, \bar{y}$  defined as in (5.9), (5.10) is independent of a rectifiable path in  $D^- \cup D$  joining two points of  $D^- \cup D$  if and only if the line integral*

$$\int L(t, x(t, \sigma), u(t, \sigma)) dt + y(t, \sigma, \rho(\sigma)) x_\sigma d\sigma \quad (5.13)$$

is independent of a rectifiable path in  $Z^- \cup Z$  joining two points in  $Z^- \cup Z$ .

**Proof.** We shall represent integral (5.1) as a line integral (5.13). By virtue of the diffeomorphism  $\theta$  (5.8), of  $D^- \cup D$  onto  $Z^- \cup Z$ , the Hilbert integral over a rectifiable curve  $\gamma$  in  $D^- \cup D$  equals a line integral of form (5.13) over the arc  $\Gamma = \theta(\gamma)$  in  $Z^- \cup Z$  provided one sets  $x = x(t, \sigma)$  in the coefficients of  $dx$  and  $dt$  in (5.1) and

$$dx = x_t(t, \sigma) dt + x_\sigma(t, \sigma) d\sigma. \quad (5.14)$$

Subject to the diffeomorphism  $\theta$  (5.8) and relations (5.11), (5.14),

$$\begin{aligned} J(\gamma) &= \int_{\gamma} \bar{y}(t, x) dx + (L(t, x, \bar{u}(t, x)) - \bar{y}(t, x) f(t, x, \bar{u}(t, x))) dt = \\ &= \int_{\Gamma} L(t, x(t, \sigma), u(t, \sigma)) dt + y(t, \sigma, \rho(\sigma)) x_\sigma d\sigma. \end{aligned} \quad (5.15)$$

We have to remember that  $\rho(\sigma)$  is defined only locally. Thus, each curve  $\Gamma$  must be divided into a finite number of arcs on which  $\rho(\sigma)$  exists and so, in (5.15) we have such sums on both sides.

Theorem 5.1 is an immediate consequence of relation (5.15). ■

Relation (5.15) has another important consequence.

**LEMMA 5.1.** *If  $\gamma$  is a subarc of an extremal trajectory of the spray  $\Sigma$ , then  $J(\gamma) = \int_{\gamma} L(t, x, u_\gamma) dt$  where  $J(\gamma)$  is defined on the left-hand side of (5.15) and  $u_\gamma$  is the control corresponding to the extremal trajectory containing  $\gamma$ .*

**Proof.** In the coordinates  $t, \sigma$  the arc  $\gamma$  has a representation  $x=x(t, \sigma)$  with  $\sigma$  constant. For this arc  $\gamma$ , one must set  $d\sigma=0$  in the right member of (5.15). If  $t \rightarrow \gamma(t)=x$  is a  $t$ -parametrization of  $\gamma$  with  $t \in [t_1, t_2] \subset [0, 1]$ , then (5.15) shows that

$$J(\gamma) = \int_{t_1}^{t_2} L(t, x(t, \sigma), u(t, \sigma)) dt.$$

Setting  $u_\gamma = u(t, \sigma)$ , we have  $J(\gamma) = \int_{t_1}^{t_2} L(t, \gamma(t), u_\gamma(t)) dt = \int_\gamma L(t, x, u) dt$ . ■

**NOTE 1.** It is clear that if  $t^+(\sigma)=1$ ,  $x(t^+(\sigma), \sigma)=e$  or  $y(t^+(\sigma), \sigma, \rho)=y^0$  for all  $\sigma \in G$  or  $(\sigma, \rho) \in \tilde{G}$ , then  $\Sigma^*$  is the canonical spray.

**NOTE 2.** If  $\Sigma^*$  is the canonical spray, then the  $\bar{u}(t, x)$ ,  $(t, x) \in [D]$ , of (5.9) is the optimal feedback control and  $\bar{y}(t, x)$ ,  $(t, x) \in [D]$  of (5.10) is the function  $S_x(t, x)$  of § 2.

**NOTE 3.** If the canonical spray  $\Sigma^*$  exists, then the function  $S(t, x)$ ,  $(t, x) \in [D]$ , (considered in §§ 1 and 2) exists and may be defined in  $[D]$  by

$$S(t, x) = - \int_t^{t^+(\sigma^*)} L(\tau, x(\tau, \sigma^*), \bar{u}(\tau, x(\tau, \sigma^*))) d\tau$$

where  $x(\tau, \sigma^*)$  is the trajectory of  $\Sigma$  passing through the point  $(t, x) \in [D]$  defined in  $[t, t^+(\sigma^*)]$  (comp. Corollary 4.1), and its exact derivative is equal to the integrand of (5.1).

**NOTE 4.** We do not make any assumptions about either continuity or even measurability of the  $u(t, \sigma)$  and, in consequence, of the feedback  $\bar{u}(t, x)$ . It is essential in practice.

The following theorem is an extension of the Weierstrass sufficiency theorem from the calculus of variations.

**THEOREM 5.2. (Sufficiency Theorem).** *Suppose that there exists a canonical spray  $\Sigma^*$  with the optimal feedback control  $\bar{u}(t, x)$ ,  $(t, x) \in [D]$ , the function  $\bar{y}(t, x)$  in  $[D]$  and the pair  $(x^0, u^0)$  of  $M$  being a member of the spray  $\Sigma$ . Then the pair  $(x^0, u^0)$  affords a minimum to  $I(x, u)$  (see (1.2)) relative to those pairs  $(x, u)$  of  $M$  whose graphs of trajectories of  $x$  lie in  $[D]$  (i.e. relative to  $M_{D\bar{u}}$ ).*

**Proof.** Denote by  $\gamma^0, \gamma$  the curves corresponding to the pairs  $(x^0, u^0)$  and  $(x, u)$  of  $M$ , where the graph of  $x$  lies in  $[D]$ , and set  $I(\gamma) = I(x, u)$ . Let  $J(\gamma)$  be the Hilbert integral defined on the left-hand side of (5.15). According to lemma 5.1,  $J(\gamma^0) =$

$=I(\gamma^0)$ . Since  $\Sigma^*$  is a canonical spray,  $J(\gamma)=J(\gamma^0)$ , so that  $I(\gamma^0)=J(\gamma)$ . Thus  $I(\gamma^0)$  is the integral (5.1) taken along  $\gamma$ . Explicitly,

$$I(\gamma^0) = \int_0^1 \left\{ L(t, x(t), \bar{u}(t, x(t))) + \bar{y}(t, x(t)) (f(t, x(t), u(t)) - f(t, x(t), \bar{u}(t, x(t)))) \right\} dt.$$

Since  $I(\gamma) = \int_0^1 L(t, x(t), u(t)) dt$ , we are led to the extension of the Weierstrass formula

$$I(\gamma) - I(\gamma^0) = \int_0^1 \left\{ H(t, x(t), \bar{u}(t, x(t)), \bar{y}(t, x(t))) - H(t, x(t), u(t), \bar{y}(t, x(t))) \right\} dt$$

where  $H(t, x, u, \bar{y})$  is defined in § 2. By virtue of the definition of a feedback control (def. 2.1.), compare also (2.6),

$$I(\gamma) - I(\gamma^0) \geq 0. \quad (5.16)$$

This completes the proof. ■

NOTE 5. If in (1.1)  $f(t, x, u) = u$  and  $U = R^n$ , then problem (1.2) becomes the standard problem from the calculus of variations and  $\bar{u}(t, x)$  is a geodesic slope or field slope,  $\bar{y}(t, x) = L_u(t, x, \bar{u}(t, x))$ , (5.16) is the integral from the Weierstrass  $E$ -condition

$$E = L(t, x, u) - L(t, x, \bar{u}) - (u - \bar{u}) L_u(t, x, \bar{u}) \geq 0, \quad (5.17)$$

and the spray  $\Sigma$  is a geodesic family or a field of extremals.

NOTE 6. In many books on the calculus of variations a field of extremals is defined as the one which induces only a line independent Hilbert integral (5.1), but then curves that define the field satisfy only the Euler equation and so, in a sufficiency theorem there must occur  $E$ -condition (5.17). In optimization theory members of  $\Sigma$  satisfy the maximum principle and thus, also a suitable inequality (5.17). Hence the last inequality does not appear in our sufficiency theorem.

Now, we give a simple example to explain the above theory.

EXAMPLE: Let  $U = [-1, 1]$ ; admissible controls are measurable functions  $u: [0, 1] \rightarrow [-1, 1]$ ; admissible trajectories are absolutely continuous functions  $x: [0, 1] \rightarrow R$  satisfying

$$\dot{x}(t) = t(u(t))^2. \quad (5.17)'$$

We find a minimum of the integral

$$I(x, u) = \int_0^1 (x(t) - (u(t))^2) dt \quad (5.18)$$

from among all admissible pairs  $(x, u)$  whose trajectories satisfy  $x(0) = 0$ ,  $x(1) = 1/2$ .

First, we calculate triplets  $(x, u, y)$  which satisfy the maximum principle. To this effect, we set  $\tilde{H}(t, x, u, y) = -L(t, x, u) + yf(t, x, u) = -x + u^2 + ytu^2$  and  $H(t, x, y) = \max_{-1 \leq u \leq 1} \tilde{H}(t, x, u, y)$ . We know (comp. 3. of § 1) that  $\dot{y} = -H_x$ . In our case this gives  $\dot{y} = 1$ . Hence  $y(t, \alpha) = t + \alpha$ . Of course,  $y(t, \alpha)$  are independent of  $x$  and  $u$ , so we can take  $\alpha = -1$ , i.e.  $y(t) = t - 1$ ,  $t \in [0, 1]$ . We easily check that  $\tilde{H}$  attains its maximum at  $u = 1$  or  $u = -1$  for each  $t \in [0, 1]$ ,  $x \in R$ . Thus  $x(t, \sigma) = \frac{1}{2}t^2 + \sigma$ ,  $t \in [0, 1]$ ,  $\sigma \in R$  and, for  $\sigma = 0$ , we have the suspected trajectory  $x^0(t) = x(t, 0) = \frac{1}{2}t^2$ .

We can now define the families  $\Sigma$  and  $\Sigma^*$ . The set  $G$  considered above is now equal to  $R$ . The functions  $t^-(\sigma)$ ,  $t^+(\sigma)$  are constant and equal to 0 and 1, respectively.  $[Z] = Z^- \cup Z \cup Z^+ = \{(0, \sigma) : \sigma \in R\} \cup \{(t, \sigma) : 0 < t < 1, \sigma \in R\} \cup \{(1, \sigma) : \sigma \in R\}$ . The family  $\Sigma$  is given by the functions

$$x(t, \sigma) = \frac{1}{2}t^2 + \sigma, \quad u(t, \sigma) = 1 \quad (t, \sigma) \in [Z]. \quad (5.19)$$

Similarly,  $[Z^*] = Z^{*-} \cup Z^* \cup Z^{*+} = \{(0, \sigma, \rho) : (\sigma, \rho) \in \tilde{G}\} \cup \{(t, \sigma, \rho) : 0 < t < 1, (\sigma, \rho) \in \tilde{G}\} \cup \{(1, \sigma, \rho) : (\sigma, \rho) \in \tilde{G}\}$  where  $\tilde{G} = G \times R = R^2$  and  $\Sigma^*$  is obtained by adjoining to functions (5.19) the conjugate function

$$y(t, \sigma, \rho) = t - 1 \quad (t, \sigma, \rho) \in [Z^*]. \quad (5.20)$$

Of course, for the function  $\rho(\sigma)$  defined in (5.4) we may take  $\rho(\sigma) = \sigma$ . It is easy to check that all assumptions (5.6) are satisfied here. Hence the diff  $\theta$  of (5.8) has

the form  $\theta: (t, x) \rightarrow (t, x - \frac{1}{2}t^2)$ , and

$$\begin{aligned} \bar{u}(t, x) &= 1 && \text{in } [D], \\ \bar{y}(t, x) &= t - 1 && \text{in } [D], \end{aligned} \quad (5.21)$$

where  $[D] = \{(t, x) : x = \frac{1}{2}t^2 + \sigma, (t, \sigma) \in [Z]\} = \{(t, x) : 0 \leq t \leq 1, x \in R\}$ . The sets  $D^-, D, D^+$  are defined analogously.

Finally, by (5.20), we obtain that integral (5.13) is equal to zero for all rectifiable paths in  $Z^+$ , so that it is independent of them there.

Thus, all assumptions of the definition of a canonical spray are satisfied, hence our family  $\Sigma^*$  is the canonical spray, and, by the sufficiency theorem, the pair  $(x^0, \bar{u})$  where  $x^0(t) = 1/2 t^2$ ,  $u^0(t) = 1$ ,  $t \in [0, 1]$ , gives the global minimum to (5.18). It is evident that the  $\bar{u}(t, x)$  of (5.21) is the optimal feedback control.

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#### **Warunki konieczne wyznaczania minimum w klasycznym zadaniu sterowania optymalnego**

Stosując klasyczny rachunek wariacyjny podano praktyczną metodę obliczania optymalnego sterowania ze sprzężeniem zwrotnym. W rezultacie otrzymano warunki dostateczne Weierstrassa wyznaczania minimum funkcjonału.

#### **Необходимые условия определения минимума в классической задаче оптимального управления**

Используя классическое вариационное исчисление, дается практический метод вычисления оптимального управления с обратной связью. В результате получены достаточные условия Вейерштрасса для определения минимума функционала.

