

OPTIMAL CONTROL PROBLEM FOR PROCESSES REPRESENTED BY STOCHASTIC SEQUENTIAL MACHINE

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ABSTRACT

In this paper, optimal control problem for processes represented by stochastic sequential machine is analyzed. Principle of optimality is proven for the considered problem. Then by using method of dynamical programming, solution of optimal control problem is found.

Keywords

Optimal control problem, stochastic sequential machine, dynamical programming.

1. INTRODUCTION

Stochastic sequential machine (SSM) is the one of the most developing field of discrete system theory [2], [3]. It plays an important role in many areas such as construction of finite dynamical system, imitation modelling problem, coding of discrete systems and identification problems. Thus, it points out that it requires a comprehensive research.

2. SSM

SSM is generalization of multi-parametric finite sequential machine[3] but it contains probability variable. General form of this system is defined by[6]:

$$K = \langle X, S, Y, s^0, p(\omega), F_v(\cdot), G \rangle \quad v = 1, 2, \dots, k \quad (1)$$

where $X = [GF(2)]^r$, $S = [GF(2)]^m$ and $Y = [GF(2)]^q$ are input, state and output index (alphabet) respectively; s^0 is initial state vector, $p(\omega)$ is deterministic discrete probability distribution ($\Omega = \{\omega_1, \omega_2, \dots, \omega_p\}$ is finite set, $p(\omega) = \{p(\omega_i) : \omega_i \in \Omega, \sum_{\omega_i \in \Omega} p(\omega_i) = 1\}$),

characteristic Boolean vector functions [7] denoted by $F_v(\cdot) = \{F_{v_1}(\cdot), \dots, F_{v_m}(\cdot)\}$ which are also known as transfer functions are nonlinear functions defined on the set $Z^k \times [GF(2)]^m \times [GF(2)]^r$ and G are an output characteristic functions defined on $GF(2)$ where $GF(2)$ is Galois field[8] and the symbol (\cdot) denotes $(c, s(c), x(c))$ for simplicity.

In addition to the definition, SSM is represented by:

$$s(c + e_v) = F_v(c, s(c), x(c), \omega(c)) \quad v = 1, 2, \dots, k \quad (2)$$

$$y(c) = G(s(c)) \quad (3)$$

where $s(c)$, $x(c)$, $y(c)$ are m , r and q dimensional state, input and output vectors at the point c respectively, $\omega(c)$ is a random variable [6], $c = \{c \mid c \in Z^k, c_1^0 \leq c_1 \leq c_1^{L_1}, \dots, c_k^0 \leq c_k \leq c_k^{L_k}, c_i \in Z\}$ is point in Z^k , determining position L_i , ($i = 1, 2, \dots, k$ positive integer) is the duration of the stage i of this process. Z is set of integers and $e_v = (0, \dots, 0, \overset{v}{1}, 0, \dots, 0)$.

In SSM, each random variable ω has a special case. For instance, ω is an input variable in the identification problem of the SSM. However, ω is a set of all possible states in the synthesis of optimal sequential machine

Moreover, the state of the system depends on the random variable ω which affects not only parameters of the SSM but also the input variable.

Finally, equation (2) is transformed to:

$$s(c + e_v) = F_v(c, s(c), x(c) \oplus \omega(c)) \quad v = 1, 2, \dots, k \quad (4)$$

where symbol \oplus means that $x \oplus w$ is always in input alphabet X .

The discrete optimal processes given by SSM are characterized by functional:

$$\bar{J}(x) = M_\omega\{\phi(s(c^L))\} \quad (5)$$

where $M_\omega\{\cdot\}$ is a mathematical expected value of ω .

3. OPTIMAL CONTROL PROBLEM AND PRINCIPLE OF OPTIMALITY

We can state optimal control problem [2] for processes represented by SSM as below: In order for given SSM to start from the known initial state s^0 to go any desired state $s^*(c^L)$, to which we expect to access in L steps, a control $x(c) \in X$ [4] must exist such that the functional in (1) has a minimal value:

$$s(c + e_v) = \hat{F}_v(c, s(c), x(c), \omega(c)), c \in G_d \quad v = 1, 2, \dots, k \quad (6)$$

$$s(c^0) = s^0, \quad x(c) \in X, c \in G_d \quad (7)$$

$$\hat{F}_v(c + e_\mu, \hat{F}_\mu(c, s(c), x(c)), x(c + e_\mu), \omega(c + e_\mu)) = \hat{F}_\mu(c + e_v, \hat{F}_v(c, s(c), x(c)), x(c + e_v), \omega(c + e_v)) \quad (8)$$

$$\bar{J}(x) = M_\omega\{\phi(s(c^L))\} \rightarrow \min \quad (9)$$

where $\hat{F}_v(\cdot)$ ($v = 1, \dots, k$) denotes the pseudo Boolean expression of the Boolean vector function [7] $\hat{F}(\cdot)$ ($v = 1, \dots, k$) and $L = L_1 + L_2 + \dots + L_k$ is the time duration of this process.

It is well-known that method of dynamic programming [5] is used for solution of optimal control problem. If we make use of this method to solve the considered problem then, (6)-(9) can be formulate as an optimal problem:

$$s(c + e_v) = \hat{F}_v(c, s(c), x(c), \omega(c)), c \in G_d(\sigma) \quad (10)$$

$$s(\sigma) = \aleph \quad (11)$$

$$x(c) \in X, c \in G_d(\sigma) \quad (12)$$

$$\hat{F}_v(c+e_\mu, \hat{F}_\mu(c, s(c), x(c), a(c)), x(c+e_\mu), a(c+e_\mu)) = \hat{F}_\mu(c+e_\nu, \hat{F}_v(c, s(c), x(c), a(c)), x(c+e_\nu), a(c+e_\nu)) \quad (13)$$

$$\bar{J}(x) = M_\omega \{ \phi(s(c^L)) \} \rightarrow \min \quad (14)$$

where \aleph is an arbitrary element in S . As it can be seen from (10)-(14), if we substitute $\sigma = c^0$ or $\aleph = s^0$ into problem (10)-(14) we obtain first problem stated above. If the conditions for existence of unique solution are satisfied then for the given initial condition $s(\sigma) = \aleph$ and given $x(c)(c \in G_d(\sigma))$, we find a unique $s(c)$. That is, the functional (14) is the function of the parameters \aleph and $x(c)(c \in G_d(\sigma))$:

$$\bar{J}(x) = \bar{J}(\aleph, x(G_d(\sigma))) \quad (15)$$

where $x(G_d(\sigma))$ denotes the range of the control $x(c)$ on the points $c \in G_d(\sigma)$:

$$x(G_d(\sigma)) = \{x(c), c \in G_d(\sigma)\} \quad (16)$$

from the unique solution condition of the system (6), we find that the stochastic process can be investigated in the set $G_d(\sigma)$ and also in the set

$$G_{d_1}(\sigma) = \{c; c_1^0 \leq c_1 < \sigma_1, \dots, c_k^0 \leq c_k < \sigma_k\} \quad (17)$$

Definition. We say that the control $x(c), (c \in G_d(\sigma))$ which minimizes the functional (5) in the problem (10)-(14) is optimal control with respect to the initial pair (σ, \aleph) on the region $G_d(\sigma)$

Suppose that $x^0(c)$ is an optimal control with respect to the initial pair (c^0, s^0) on the region G_d and $s^0(c)$ is admissible optimal trajectory. Then $x^0(c)$ is an optimal control with respect to the initial pair $(\sigma, s^0(\sigma))$ on the region $G_d(\sigma)$ for every G_d .

Proof. Suppose the contrary. Then there exists $x(c) \in X, c \in G_d(\sigma)$ such that we have

$$\bar{J}(\aleph, x(G_d(\sigma))) < \bar{J}(\aleph, x^0(G_d(\sigma))) \quad (18)$$

We choose a new control process $\tilde{x}(c), c \in G_d$ as follows:

$$\tilde{x}(c) = \begin{cases} x^0(c), c \in G_{d_1}(\sigma) \\ x(c), c \in G_d(\sigma) \end{cases} \quad (19)$$

As it can be seen, (19) is an admissible control such that

$$\bar{J}(s^0, \tilde{x}(G_d)) = \bar{J}(s^0, \tilde{x}(G_{d_1}(\sigma) \cup G_d(\sigma))). \quad (20)$$

According to the condition, $s^0(\sigma) = \mathfrak{K}$. Thus we have

$$\begin{aligned} \bar{J}(s^0, \tilde{x}(G_{d_1}(\sigma) \cup G_d(\sigma))) &= \bar{J}(s^0(\sigma), \tilde{x}(G_d(\sigma))) = \bar{J}(\mathfrak{K}, x(G_d(\sigma))) < \bar{J}(\mathfrak{K}, x^0(G_d(\sigma))) = \\ &= \bar{J}(s(\sigma), x^0(G_d(\sigma))) = \bar{J}(s^0, x^0(G_d)) \end{aligned} \quad (21)$$

and by virtue of (2.10) and (2.11) we can obtain

$$\bar{J}(s^0, \tilde{x}(G_d)) < \bar{J}(s^0, x^0(G_d)) \quad (22)$$

Hence, (22) contradicts the hypothesis that the control $x^0(c), c \in G_d$ is optimal. This completes the proof of the theorem.

Let a function (for every fixed σ and \mathfrak{K}) be corresponding to the optimal value of pseudo Boolean functional in the problem (10)-(14):

$$B(\sigma, \mathfrak{K}) = \min M_{\omega} \{ \phi(s(c^L)) \}. \quad (23)$$

where minimization on the set of admissible control $x(c), c \in G_d(\sigma)$.

Now, we determine method of dynamical programming (It is known as Bellman equation) [5] for $B(\sigma, \mathfrak{K})$. Suppose that $x^0(c), c \in G_d$ is the admissible control corresponding to (10)-(14) with initial condition and $s^0(c), c \in G_d(\sigma)$ is also the optimal trajectory. Let the point $\xi_v, s \in G_d(\sigma) (v=1,2,\dots,k)$ and any element $y(c) \in X$ be specified. If $x(\sigma) = y(c)$, then the state of the system in the point $\xi_v \sigma$ is determined by

$$s(\xi_v \sigma) = \hat{F}_v(\sigma, \mathfrak{K}, y, \omega(\sigma)) \quad (24)$$

We consider the following problem:

$$\xi_v s(c) = \hat{F}_v(c, s(c), x(c), \omega(c)), c \in G_d(\xi_v \sigma) \quad (25)$$

$$s(\xi_v \sigma) = \hat{F}_v(\sigma, \mathfrak{K}, y) \quad (26)$$

$$x(c) \in X(c), c \in G_d(\xi_v \sigma) \quad (27)$$

$$\bar{J}(x) = M_{\omega} \{ \phi(s(c^L)) \} \rightarrow \min \quad (28)$$

If $\hat{y}(c), c \in G_d(\xi_v \sigma)$ and $\hat{s}(c), c \in G_d(\xi_v \sigma)$ are optimal control and optimal trajectory respectively, then

$$M_{\omega} \{ \phi(s(c^L)) \} = B(\xi_v \sigma, \hat{F}_v(\sigma, \mathfrak{K}, \hat{y}, \omega(\sigma))) \quad (29)$$

can be found.

For (10)-(14), let $\tilde{x}(c)$ be an admissible control below.

$$\tilde{x}(c) = \begin{cases} y, & c = \sigma \\ \hat{y}(c), & c \in G_d(\xi_v \sigma) \end{cases} \quad (30)$$

Also, $\tilde{s}(c)$ can be obtained by

$$\tilde{s}(c) = \begin{cases} \hat{F}_v(\sigma, \mathfrak{K}, y, \omega(\sigma)), & c = \sigma \\ \hat{s}(c), & c \in G_d(\xi_v \sigma) \end{cases} \quad (31)$$

It is evident that the value of $\bar{J}(x) = M_{\omega} \{ \phi(s(c^L)) \}$ to control $\tilde{x}(c)$ is determined by

$$M_{\omega} \{ \phi(\tilde{s}(c^L)) \} = M_{\omega} \{ \phi(\hat{s}(c^L)) \} = B(\xi_v \sigma, \hat{F}_v(\sigma, \mathfrak{K}, \hat{y}, \omega(\sigma))). \quad (32)$$

Since $\tilde{x}(c), c \in G_d(\sigma)$ is not largely optimal control, we can state

$$M_{\omega} \{ \phi(\tilde{s}(c^L)) \} \geq M_{\omega} \{ \phi(s^0(c^L)) \} = B(\sigma, \mathfrak{K}) \quad (33)$$

Thus, we have

$$B(\sigma, \mathfrak{K}) \leq B(\xi_v \sigma, \hat{F}_v(\sigma, \mathfrak{K}, \hat{y}, \omega(\sigma))) \quad (34)$$

On the other hand, if $y(c) = x^0(\sigma)$, then by the principle of optimality [2],

$$\hat{y}(c)(c \in G_d(\xi_v \sigma)) = x^0(c)(c \in G_d(\xi_v \sigma)) \quad (35)$$

Therefore,

$$B(\sigma, \mathfrak{K}) = B(\xi_v \sigma, \hat{F}_v(\sigma, \mathfrak{K}, x^0(\sigma), \omega(\sigma))) \quad (36)$$

By (34) and (35), Bellman equation[5] can be determined by

$$B(\sigma, \mathfrak{K}) = \min_{y \in X(\sigma)} B(\xi_v \sigma, \hat{F}_v(\sigma, \mathfrak{K}, \hat{y}, \omega(\sigma))), \quad (37)$$

$$\mathfrak{K} \in S$$

4. CONCLUSIONS

It is shown that Bellman equation for optimal processes with stochastic sequential machine is obtained and the principle of optimality is proven.

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