BELLMAN EQUATION FOR OPTIMAL PROCESSES WITH NONLINEAR MULTI-PARAMETRIC BINARY DYNAMIC SYSTEM

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In general, nonlinear multi-parametric binary dynamic system(NMBDS) is defined as follows [1]:

$$\xi_{v}s(c) = F_{v}(c, s(c), x(c)), v = 1, 2, ..., k$$
(1)

$$s(c^0) = s^0 \tag{2}$$

where $c = (c_1, c_2, ..., c_k) \in G_d = \{c \mid c \in Z^k, c_1^0 \leq c_1 \leq c_1^{L_1}, ..., c_k^0 \leq c_k \leq c_k^{L_k}, c_i \in Z\}$ is a point in Z^k determining position; $L_i, i = 1, 2, ..., k$ where k is a positive integer, is the duration of the stage *i* of the process. Here, *Z* is the set of integers. For $s(c) \in S, x(c) \in X; S = [GF(2)]^m$, $X = [GF(2)]^r$ are state and input index (alphabet) respectively; s(c) and x(c) are defined over the set Z^k as an *m* and *r* dimensional state and input vectors at the point *c*. $c^0 = (c_1^0, c_2^0, ..., c_k^0)$ is the initial position vector of the system and s^0 is the initial state vector of the system. $c^{L_i} = (c_1^{L_i}, c_2^{L_i}, ..., c_k^{L_i})$ is the point to which the system moves after the stage

i-1. ξ_v is a shift operator defined as follows [1]:

$$\xi_{v}s(c) = s(c+e_{v}); e_{v}(0,...,0,1,0,...,0), v = 1,2,...,k.$$
(3)

Boolean vector functions [2,3] denoted by $F_{v}(\cdot) = \{F_{v_1}(\cdot), F_{v_2}(\cdot), \dots, F_{v_m}(\cdot)\}$ are nonlinear functions, where GF(2) is a Galois field and the representation (\cdot) denotes (c, s(c), x(c)) for simplicity.

Optimal piecewise process represented by the system (1)-(2) is characterized by the pseudo Boolean functional [3] given by:

$$J(x) = \varphi(s(c^{L})) \tag{4}$$

which we use as an objective functional for the considered problem in the presented proceeding. Here, $L = L_1 + L_2 + ... + L_k$ is the time duration of this process.

Now, we can state the considered original problem represented by NMBDS as follows: In order for a given NMBDS to go from the known initial state s^0 to any desired state $s^*(c^L)$,

to which we expect to access in *L* steps, a control $x(c) \in X$ [6] must exist such that the functional in (4) has a minimal value:

$$\xi_{v}s(c) = F_{v}(c, s(c), x(c)), \ c \in G_{d}, \ v = 1, 2, ..., k$$
(5)

$$s(c^0) = s^0 \tag{6}$$

$$x(c) \in \overset{\frown}{X}, \ c \in \overset{\frown}{G}_d \tag{7}$$

$$J(x) = \varphi(s(c^{L})) \to \min.$$
(8)

Since the transfer functions are Boolean and the objective functional which characterizes the process is the process is pseudo Boolean [3,5], the pseudo Boolean expressions of the transfer functions have been obtained by the operations given in [5]. After this step, the problem can be stated as follows:

$$\xi_{v}s(c) = \hat{F}_{v}(c, s(c), x(c)), \ c \in G_{d}, \ v = 1, 2, ..., k$$
(9)

$$s(c^0) = s^0 \tag{10}$$

$$x(c) \in \overset{\circ}{X}, c \in \overset{\circ}{G}_{d} \tag{11}$$

$$J(x) = \varphi(s(c^{L})) \to \min.$$
⁽¹²⁾

where $\hat{F}_{v}(\cdot)(v=1,...,k)$ denotes the pseudo Boolean expression of the Boolean vector function

 $F_{v}(\cdot)(v=1,...,k)$ and $G_{d} = G_{d} \setminus \{c^{L}\}.$

Since we have shown that the principle of optimality [4] is satisfied for the problem (9)-(12), hereafter we can formulate this problem as an optimal problem [6]:

$$\xi_{v}s(c) = F_{v}(c, s(c), x(c)), \ c \in G_{d}(\sigma), \ v = 1, 2, ..., k$$
(13)

$$s(\sigma) = \chi \tag{14}$$

$$\mathbf{x}(c) \in \widehat{X}, \ c \in G_d(\sigma) \tag{15}$$

$$J(x) = \varphi(s(c^{L})) \to \min$$
(16)

where $\chi \in S = [GF(2)]^m$, $\sigma \in G_d$, $G_d(\sigma) = \{c | \sigma_1 \le c_1 \le c_1^{L_1}, ..., | \sigma_k \le c_k \le c_k^{L_k}\}$. If we substitute $\sigma = c^0$ and $\chi = s^0$ into the problem (13)-(16), we can obtain the first problem stated above.

For every fixed σ and χ , let a function be corresponded to the optimal value of pseudo Boolean functional in the problem (13)-(16). We say that this function is the piecewise analogue of Bellman function [4] in the problem (9)-(12):

$$B(\sigma, \chi) = \min \varphi(s(c^{L})).$$
(17)

Here, minimization is implemented on the set of admissible controls x(c) ($c \in G_d(\sigma)$).

We derive the Bellman equation for $B(\sigma, \chi)$ function: Assume that $x^0(c)$ ($c \in G_d(\sigma)$) is the corresponding optimal control to the problem (13)-(16) with the initial condition and $s^0(c)$ ($c \in G_d(\sigma)$) is also the corresponding optimal trajectory to that problem.

Let the point $\xi_v \sigma \in G_d(\sigma)$ (v = 1, 2, ..., k) and any element $y(c) \in \hat{X}$ be considered. If $x(\sigma) = y(c)$, then the state of the system in the point $\xi_v \sigma$ is determined by the following equality:

$$s(\xi_{v}\sigma) = F_{v}(\sigma, \chi, y(c)).$$
⁽¹⁸⁾

We consider the following problem:

$$\xi_{v}s(c) = F_{v}(c, s(c), x(c)), \ c \in G_{d}(\xi_{v}\sigma)$$
⁽¹⁹⁾

$$s(\xi_v \sigma) = F_v(\sigma, \chi, y(c)) \tag{20}$$

$$x(c) \in \overline{X}, c \in G_d(\xi_v \sigma) \tag{21}$$

$$J(x) = \varphi(s(c^{L})) \to \min.$$
⁽²²⁾

If y(c) $(c \in G_d(\xi_v \sigma))$ is the corresponding optimal control to the problem (19)-(22) and $\hat{s}(c)$ $(c \in G_d(\xi_v \sigma))$ is also the corresponding optimal trajectory to that problem, then according to our definition stated above, the equality

$$\varphi(s(c^{L})) = B(\xi_{v}\sigma, F_{v}(\sigma, \chi, y(c)))$$
⁽²³⁾

can be obtained.

Now, let the following admissible control

$$\widetilde{x}(c) = \begin{cases} y(c), & \text{for } c = \sigma \\ & \\ y(c), & \text{for } c \in G_d(\xi_v \sigma) \end{cases}$$
(24)

be considered for the problem (13)-(16). Then $\tilde{s}(c)$ is determined by

$$\widetilde{s}(c) = \begin{cases} F_{v}(\sigma, \chi, y(c)), & \text{for } c = \sigma \\ & \hat{s}(c), & \text{for } c \in G_{d}(\xi_{v}\sigma) \end{cases}$$
(25)

It is clear that the corresponding value of the pseudo Boolean functional $J(x) = \varphi(s(c^{L}))$ to the control $\hat{x}(c)$ ($c \in G_{d}(\sigma)$) is

$$\varphi(\widetilde{s}(c^{L})) = \varphi(\widetilde{s}(c^{L})) = B(\xi_{v}\sigma, F_{v}(\sigma, \chi, y(c))) \quad .$$
⁽²⁶⁾

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

$$\varphi(\tilde{s}(c^{L})) \ge \varphi(s^{0}(c^{L})) = B(\sigma, \chi) \quad .$$
⁽²⁷⁾

Thus, we have

$$B(\sigma, \chi) \le B(\xi_{\nu}\sigma, F_{\nu}(\sigma, \chi, y(c))) \quad .$$
⁽²⁸⁾

On the other hand, if $y(c) = x^0(\sigma)$, then $y(c) = (c \in G_d(\xi_v \sigma))$ equals to $x^0(c)$ ($c \in G_d(\xi_v \sigma)$) by the principle of optimality. So,

$$B(\sigma, \chi) = B(\xi_{\nu}\sigma, F_{\nu}(\sigma, \chi, x^{0}(\sigma))) \quad .$$
⁽²⁹⁾

By (28) and (29), Bellman equation can be obtained as follows:

$$B(\sigma, \chi) = \min_{\substack{y(c) \in X}} B(\xi_{\nu}\sigma, F_{\nu}(\sigma, \chi, x^{0}(\sigma))), \ \chi \in S.$$
(30)

The initial condition for Bellman equation is given on the right-upper region of G_d and directly determined with the help of the following equality

$$B(c^{L},\chi) = \varphi(\chi), \ \chi \in S \ . \tag{31}$$

Hence, Bellman function is the solution of the equation (30) with the initial condition (31). It is clear that Bellman equation (30) has exact solution.

If η_v is inverse operator of the shift operator ξ_v , then we have

$$\eta_{\nu}B(\xi_{\nu}\sigma,\chi) = B(\sigma,\chi).$$
(32)

Thus, Bellman equation (30) can be derived as follows:

$$\eta_{v}B(\xi_{v}\sigma,\chi) = \min_{y(c)\in\hat{X}} B(\xi_{v}\sigma,F_{v}(\sigma,\chi,y(c))), v = 1,2,...,k.$$
(33)

Substituting $\xi_v \sigma = \delta$ in (33), we obtain

$$\eta_{\nu}B(\delta,\chi) = \min_{\substack{y(c)\in \hat{X}(\eta_{\nu}\delta)}} B(\delta, F_{\nu}(\eta_{\nu}\delta,\chi,y(c))), \nu = 1,2,...,k.$$
(34)

If $\eta_v(\eta_{v'}B(\delta,\chi))$ is evaluated for every v, v' = 1, 2, ..., k, then

$$\eta_{v}(\eta_{v'}B(\delta,\chi)) = \eta_{v}(\min_{y(c)\in\hat{X}(\eta_{v'}\delta)} B(\delta, F_{v'}(\eta_{v'}\delta,\chi,y(c))))$$
$$= \min_{y(c)\in\hat{X}(\eta_{v'}\delta)} \eta_{v}B(\delta, F_{v'}(\eta_{v'}\delta,\chi,y(c)))$$

$$= \min_{\substack{\gamma(c)\in X (\eta,\eta_{v}\delta)}} \left[\min_{x(c)\in X (\eta_{v}\delta)} B(\delta, F_{v}(\eta_{v}\delta, F_{v'}(\eta_{v'}\eta_{v}\delta, y(c), \chi), x(c))) \right].$$
(35)

Similarly,

$$\eta_{v'}(\eta_{v}B(\delta,\chi)) = \min_{\substack{y(c)\in \hat{X}(\eta_{v},\eta_{v}\delta) \ x(c)\in \hat{X}(\eta_{v'}\delta)}} \left[\min_{\substack{x(c)\in \hat{X}(\eta_{v'}\delta) \ x(c)\in \hat{X}(\eta_{v'}\delta)}} B(\delta, F_{v'}(\eta_{v'}\delta, F_{v}(\eta_{v}\eta_{v'}\delta, y(c),\chi), x(c)))\right].$$
(36)

The condition implying the existence of the unique solution of the system of equations (13) is given by [1]

$$F_{\nu}(c + e_{\mu}, F_{\mu}(c, s(c), x(c)), x(c + e_{\mu})) = F_{\mu}(c + e_{\nu}, F_{\nu}(c, s(c), x(c)), x(c + e_{\nu})),$$

$$v, \mu = 1, 2, ..., k.$$
(37)

Substituting $\mu = v'$ in (37), we obtain

$$\hat{F}_{\nu}(\eta_{\nu}\delta, \hat{F}_{\nu'}(\eta_{\nu}\eta_{\nu'}\delta, \chi, x(\eta_{\nu}\eta_{\nu'}\delta)), x(\eta_{\nu}\delta)) =$$

$$= \hat{F}_{\nu'}(\eta_{\nu'}\delta, \hat{F}_{\nu}(\eta_{\nu'}\eta_{\nu}\delta, \chi, x(\eta_{\nu'}\eta_{\nu}\delta)), x(\eta_{\nu'}\delta)), \nu, \nu' = 1, 2, ..., k.$$
(38)

By (35), (36) and (38), the following result is derived:

 $\eta_{\nu}(\eta_{\nu'}B(\delta,\chi)) = \eta_{\nu'}(\eta_{\nu}B(\delta,\chi)), \nu,\nu' = 1,2,...,k.$

This result is the condition implying the existence of the exact solution for Bellman equation (30). If Bellman equation (30) is solved subject to the condition (31) over the curve $\hat{I}(c^0, c^1, c^L)$ then we achieve the following functions after *L* steps:

$$(c^{\circ}, c^{*}, ..., c^{2})$$
, then we achieve the following functions after L steps

$$B(c^{L},\chi), B(c^{L-1},\chi),...,B(c^{0},\chi).$$

 $B(c^0, s(c^0))$ is the minimal value of the pseudo Boolean functional in the problem (13)-(16). The optimal control is determined with the help of the following condition:

$$B(\xi_{v}c, F_{v}(c, s(c), x^{0}(c))) = \min_{x(c) \in \hat{X}} B(\xi_{v}c, F_{v}(c, s(c), x(c))).$$

where $c \in \hat{L}(c^0, c^1, ..., c^L)$. Here, $\hat{L}(c^0, c^1, ..., c^L)$ is piecewise curve associating the point c^0 with the point $c^L[1]$. v takes value such that $\xi_v c \in \hat{L}(c^0, c^1, ..., c^L)$. Then, the optimal trajectory is determined by

$$\xi_{v}s(c) = \hat{F_{v}}(c, s(c), x^{0}(c)), v = 1, 2, ..., k$$
$$s(c^{0}) = s^{0}.$$

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