

ON THE PROBLEM OF CONVERGENCE OF COSTS

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1. The problem of convergence of costs in the Kalman–Bucy scheme of partially observable random processes has been studied in a lot of works (see, for example, [3], [4], [5]). In the present paper, this problem is being studied for the generalized Kalman–Bucy scheme ([2]).

Consider the generalized Kalman–Bucy scheme of partially observable random processes ([2])

$$Y_t = Y_0 + Y \circ A_1(t) + M_1(t), \quad t \geq 0, \quad (1)$$

$$X_t = X_0 + Y \circ A_2(t) + \varepsilon M_2(t), \quad t \geq 0, \quad (2)$$

where $M_1(t)$ and $M_2(t)$ are the local martingales, $A_1(t)$ and $A_2(t)$ are the deterministic functions, and $(Y_0, X_0, M_1(t), M_2(t))$ is the Gaussian system. $H \circ A$ denotes the Lebesgue–Stieltjes integral of the process H with respect to the process A .

Assume that the gain function is linear and has the form

$$g(t, x) = f_0(t) + f_1(t)x, \quad (3)$$

where $f_0(t)$ and $f_1(t)$ are the deterministic measurable functions.

Introducer the costs

$$s^0 = \sup_{\tau \in \mathfrak{M}^Y} Eg(\tau, Y_\tau), \quad (4)$$

$$s^\varepsilon = \sup_{\tau \in \mathfrak{M}^X} Eg(\tau, Y_\tau), \quad (5)$$

where \mathfrak{M}^y and \mathfrak{M}^X are the classes of stopping moments with respect to the families of σ -algebras (\mathcal{F}_t^Y) and (\mathcal{F}_t^X) , $\mathcal{F}_t^Y = \sigma\{Y_s, 0 \leq s \leq t\}$, $\mathcal{F}_t^X = \sigma\{X_s, 0 \leq s \leq t\}$.

Suppose now that for the functions $g_0(t)$, $g_1(t)$, $g_2(t)$ and for the increasing process Z_t the following conditions are fulfilled:

- I. $EY^2g^2 \circ \langle M_2 \rangle$ is increasing;
- II. $\langle M_1 \rangle = g_1 \circ Z$, $\langle M_2 \rangle = g_2 \circ Z$;
- III. $\Delta A_2 = 0$, $\langle M_1, M_2 \rangle = 0$,

where $\langle M_1 \rangle$ and $\langle M_2 \rangle$ are quadratic characteristics of the local martingales M_1 and M_2 , and $\langle M_1, M_2 \rangle$ is their mutual quadratic characteristic.

Further, let $\mathcal{E}_t(A_1)$ denote a stochastic exponent, or a solution of the linear stochastic Dolean equation

$$\mathcal{E}_t(A_1) = 1 + \mathcal{E}_t(A_1) \circ A_1, \quad (6)$$

and let $\rho(t)$ be a continuous increasing function, majorizing the function $(1 + \Delta A_1)^2 g_2 |g_0^2$. Let, moreover,

$$m_t = E(Y | \mathcal{F}^X), \quad \gamma_t = E(Y_t - m_t)^2. \quad (7)$$

2. In the theorem below we will prove the convergence of the cost s^ε to the cost s^0 , as $\varepsilon \rightarrow 0$. Assume that $0 \leq f_1(t) \leq F < \infty$.

Theorem. *Let a partially observable random process be given by the equations (1), (2), the costs (4), (5) be defined, and the conditions I–III be fulfilled. Then the estimate*

$$s^0 - s^\varepsilon \leq \varepsilon \cdot F \cdot \rho(t) \cdot \mathcal{E}_t^2(A_1) \quad (8)$$

holds.

proof. First of all, it should be noted that the difference of costs can be estimated by γ_t as follows ([4]):

$$s^0 - s^\varepsilon \leq F \cdot \gamma_t, \quad (9)$$

where γ_t satisfies the equation

$$\begin{aligned} \gamma_t &= \gamma_0 + \gamma_t(2 + \Delta A_1) \circ A_2 + \langle M_1 \rangle - q^2 \circ (\langle \varepsilon M_2 \rangle), \\ q &= \frac{d[\gamma_t \cdot (1 - \Delta A_1) \circ A_2]}{d(\langle \varepsilon M_2 \rangle)}. \end{aligned}$$

Introduce the transformation

$$\gamma_t = \varepsilon \cdot u_t \cdot \mathcal{E}_t^2(A_1) \quad (10)$$

and show that $u_t \leq \rho(t)$ for every $t \geq 0$. Thus the theorem is complete. We have

$$\begin{aligned} \gamma_t &= \int_0^t \gamma_s \cdot (2 + \Delta A_1) g_1(s) dZ_s + \\ &+ \int_0^t g_1(s) dZ_s - \frac{1}{\varepsilon^2} \int_0^t \frac{\gamma_s^2 \cdot (1 + \Delta A_1)^2 g^2(s)}{g_2(s)} dZ_s, \end{aligned}$$

whence for u_t we can write

$$\begin{aligned} \varepsilon \int_0^t \mathcal{E}_s^2(A_1) du_s &= \int_0^t g_1(s) dZ_s - \\ &- \int_0^t \frac{u_s^2 \cdot \mathcal{E}_s^4(A_1) (1 + \Delta A_1)^2 g^2(s)}{g_2(s)} dZ_s. \end{aligned}$$

From the last relation it immediately follows that

$$\begin{aligned} u_t &= \frac{1}{\varepsilon} \int_0^t \mathcal{E}_s^2(A_1) g_1(s) dZ_s - \frac{1}{\varepsilon} \int_0^t \frac{u_s^2 \cdot \mathcal{E}_s^2(A_1) g^2(s)}{g_2(s)} dZ_s = \\ &= \frac{1}{\varepsilon} \int_0^t \frac{\mathcal{E}_s^2(A_1) g^2(s)}{g_2(s)} \left[\frac{(1 + \Delta A_1)^2 g_2(s)}{g^2(s)} - u_s^2 \right] dZ_s. \end{aligned} \quad (11)$$

Just as in [4], from (11) we can conclude that $u_t \leq \rho(t)$, which provides us with the estimate (8).

R E F E R E N C E S

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