

9. G_9 -ESTIMATOR OF THE SOLUTION OF THE DISCRETE KOLMOGOROV-WIENER FILTER

The discrete analog of the Kolmogorov-Wiener filter has the form

$$R_m \vec{\varphi} = \vec{b}, \quad (9.1)$$

where

$$R_m = \{m^{-1}R(sm^{-1}, km^{-1})\}_{k,s=1}^m; \quad \vec{b}^T(t) = \{Q(t, sm^{-1}), s = 1, \dots, m\},$$

$$\vec{\varphi}^T(t) = \{\varphi(t, km^{-1}), k = 1, \dots, m\},$$

$$R(x, y) = \mathbf{E} [\alpha(x) - \mathbf{E} \alpha(x)] [\alpha(y) - \mathbf{E} \alpha(y)],$$

$$Q(x, y) = \mathbf{E} [\alpha(x) - \mathbf{E} \alpha(x)] [\beta(y) - \mathbf{E} \beta(y)],$$

and $\alpha(x)$, $\beta(y)$ are random processes. If $R_m > 0$, then the estimator $\hat{\vec{\varphi}} = (\hat{R}_m)^{-1} \hat{\vec{b}}$ converges in probability to $\vec{\varphi}$ when $n_1, n_2 \rightarrow \infty$, where

$$\hat{R} = \left\{ m^{-1} \hat{R}(sm^{-1}, km^{-1}) \right\}_{k,s=1}^m, \quad \vec{\varphi}^T(t) = \{\varphi(t, km^{-1}), k = 1, \dots, m\};$$

$$\hat{\vec{b}}^T(t) = \{\hat{Q}(t, sm^{-1}), s = 1, \dots, m\},$$

$$\hat{R}(x, y) = (n_1 - 1)^{-1} \sum_{k=1}^{n_1} [\alpha_k(x) - \hat{\alpha}(x)] [\alpha_k(y) - \hat{\alpha}(y)],$$

$$\hat{Q}(x, y) = (n_2 - 1)^{-1} \sum_{k=1}^{n_2} [\alpha_k(x) - \hat{\alpha}(x)] [\beta_k(y) - \hat{\beta}(y)],$$

and $\alpha_k(x)$, $\beta_k(y)$ are independent observations of $\alpha(x)$, $\beta(y)$.

As mentioned in previous sections of this chapter, the large order of system (9.1) requires a large number of observations of stochastic processes $\alpha(x)$, $\beta(y)$.

Therefore, it is of interest to obtain more accurate estimators. Applying the G -analysis technique, which is described in [Gir44, Gir54, Gir69, Gir84], we can obtain an estimator of $\vec{\varphi}$, such that it would approach in probability $\vec{\varphi}$, provided that $\lim_{n \rightarrow \infty} mn^{-1} = c < 1$. We assume for simplification of formulas that vector \vec{b} is known. This estimator will be referred to as the G_9 -estimator. It is

$$\vec{G}_9 = (\hat{R}_m)^{-1} \vec{b} \left(1 - \frac{m_n}{n}\right). \quad (9.2)$$

Denote

$$\vec{\alpha}_k^T = \left(\alpha_k \left(\frac{s}{m} \right), s = 1, \dots, m \right), \quad R^{-1/2}(\vec{\alpha}_k - \mathbf{E} \vec{\alpha}_k) = \vec{\xi}_k = (\xi_{sk} \ s = 1, \dots, m)^T.$$

THEOREM 9.1. ([Gir44], [Gir54], [Gir69], [Gir84]) *If random variables ξ_{sk} are independent for every n , $\mathbf{E}|\xi_{ki}|^{4+\delta} \leq c$, $\delta > 0$, $\lim_{n_1 \rightarrow \infty} mn_1^{-1} < 1$; $\lambda_i(R_m) \leq c < \infty$, the vector \vec{b} is known,*

$$\sup_m [\vec{b}^T \vec{b} + \vec{c}^T \vec{c}] < \infty,$$

where $\vec{c} \in R^m$, and $\lambda_i(R_m)$ are the eigenvalues of the matrix R_m , then

$$p \lim_{n_1 \rightarrow \infty} [\vec{c}^T \vec{G}_9 - \vec{c}^T \vec{\varphi}] = 0.$$

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