Linear Filtration Methods for Statistical Analysis of Periodically Correlated Random Processes

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Abstract - Coherent and component methods for mean and covariance function estimation are analyzed using linear filtration theory.

Keywords - Cyclostationary Processes, Coherent Method, Component Method, Linear Filtration Methods.

I. INTRODUCTION

Periodically correlated (PC) random processes are popular models of physical phenomena with cyclic and stochastic behavior [1, 2]. They are also known as cyclostationary or periodic stationary random processes. This means in particular that mean function of real-valued process m(t) = Ex(t) = m(t+T), covariation function $b(t,u) = E\overline{x}(t)\overline{x}(t+T,u) = b(t+T,u), \overline{x}(t) = x(t) - m(t)$. Also we may use the following Fourier series representation

$$m(t) = \sum_{k \in \mathbf{c}} m_k e^{ikw_0 t}, \ b(t, u) = \sum_{k \in \mathbf{c}} B_k(u) e^{ikw_0}$$

Traditional methods for PC processes statistical analysis are coherent and component methods. The coherent method in estimating first and second order characteristics of PC signals are based on synchronous averaging:

$$\hat{m}(t) = \frac{1}{N} \sum_{n=0}^{N-1} \mathbf{x}(t+nT),$$
$$\hat{b}(t,u) = \frac{1}{N} \sum_{n=0}^{N-1} [\mathbf{x}(t+nT) - \hat{m}(t+nT)] \times \mathbf{x}[\mathbf{x}(t+u+nT) - \hat{m}(t+u+nT)],$$

here N is a number of periods, that are averaged, hereby the length of realization fragment $q = NT + u_m$, where u_m is the autocovariance function maximal time lag. Component estimates are built like trigonometric polynomial:

$$\hat{m}(t) = \sum_{k=-N_1}^{N_1} \hat{m}_k e^{ikw_0 t}, \quad \hat{b}(t,u) = \sum_{k=-N_2}^{N_2} \hat{B}_k(u) e^{ikw_0 t},$$

where

$$\hat{m}_{k} = \frac{1}{q} \int_{0}^{q} \mathbf{X}(t) e^{-ikw_{0}t} dt,$$
$$\hat{B}_{k}(u) = \frac{1}{q} \int_{0}^{q} [\mathbf{X}(t) - \hat{m}(t)] [\mathbf{X}(t+u) - \hat{m}(t+u)] e^{-ikw_{0}t} dt$$

More detailed analysis of coherent and component estimates can be found in [1]. In this paper we analyze these methods from the point of view of linear filtration theory.

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II. LINEAR FILTRATION IN PC SIGNALS

Coherent and component methods for probability characteristic estimation of PC process can be represented as particular cases of demodulation method [1], there for the estimator $\hat{m}(t)$ we have:

$$\hat{m}(t) = \int_{t_0}^t x(t)h(t-t)dt = \int_0^{t-t_0} x(t-t)h(t)dt,$$

where h(t) is the weight function. Here we suppose that signal is given in the interval $[t_0, t]$ and its length $q = t - t_0 = NT$.

If the weight function h(t) is taken in following form:

$$h(t) = \frac{1}{N} \sum_{n=0}^{N-1} d(t - nT),$$

then we obtain a coherent estimate:

$$\hat{m}(t) = \frac{1}{N} \sum_{n=0}^{N-1} \int_{0}^{t-t_{0}} x(t-t) d(t-nT) dt = \frac{1}{N} \sum_{n=0}^{N-1} x(t-nT).$$

When weight function has a form

$$h(t) = \frac{1}{t - t_0} \sum_{k = -N_1}^{N_1} e^{ikw_0 t},$$

we obtain the component estimate:

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$$\hat{h}(t) = \int_{0}^{t-t_{0}} \mathbf{x}(t-t) \left[\frac{1}{t-t_{0}} \sum_{k=-N_{1}}^{N_{1}} e^{ikw_{0}t}\right] dt =$$
$$= \sum_{k=-N_{1}}^{N_{1}} e^{ikw_{0}t} \left[\frac{1}{t-t_{0}} \int_{t_{0}}^{t} \mathbf{x}(t) e^{ikw_{0}t} dt\right].$$

Transfer function of the filter in the case of coherent estimate has the form

$$H(W) = \frac{1}{N} \sum_{n=0}^{N-1} e^{-iwnT} = e^{i\frac{WT}{2}} (N-1) \frac{\sin[\frac{WWT}{2}]}{N \sin[\frac{WT}{2}]},$$

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and in the case of component estimate

$$H(w) = \sum_{k=-N_1}^{N_1} e^{-i(w-kw_0)\frac{q}{2}} \sin[(w-kw_0)\frac{q}{2}] / (w-kw_0)\frac{q}{2}$$

III. CONCLUSION

General approach for PC process estimation theory can be developed on the base of linear filtration methods.

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