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Gauss-Markov whitening, Generalized Least Squares

Noise covariance matrix

$$\text{cov}(Z_N) = \mathbf{E}Z_N Z_N^T$$

is positive definite. We can apply factorization theorem

Twierdzenie 1 *Each symmetric and positive definite matrix \mathbf{M} can be written in the form*

$$\mathbf{M} = \mathbf{P}\mathbf{P}^T$$

where \mathbf{P} is nonsingular (called "root of \mathbf{M} ").

One can write that

$$\text{cov}(Z_N) = P \cdot P^T = C$$

apply P^{-1} for measurement equation

$$P^{-1}Y_N = P^{-1}\Phi_N\theta + P^{-1}Z_N$$

Introducing

$$\bar{Y}_N = P^{-1}Y_N, \quad \bar{\Phi}_N = P^{-1}\Phi_N, \quad \bar{Z}_N = P^{-1}Z_N,$$

we get

$$\bar{Y}_N = \bar{\Phi}_N\theta + \bar{Z}_N$$

What is whitening?

$$\mathbf{E}\bar{Z}_N\bar{Z}_N^T = \mathbf{E}P^{-1}Z_N Z_N^T P^{-1T} = P^{-1}(\mathbf{E}Z_N Z_N^T)P^{-1T} = P^{-1}P \cdot P^T P^{-1T} = I$$

$$\mathbf{E}\bar{Z}_N\bar{Z}_N^T = I \quad (I - \text{identity matrix, diagonal} - \bar{Z} \text{ is white}),$$

GLS method

$$\widehat{\theta}_{GLS} = (\bar{\Phi}_N^T \bar{\Phi}_N)^{-1} \bar{\Phi}_N^T \bar{Y}_N$$

$$\widehat{\theta}_{GLS} = (\Phi_N^T P^{-1T} P^{-1} \Phi_N)^{-1} \Phi_N^T P^{-1T} P^{-1} Y_N = (\Phi_N^T C^{-1} \Phi_N)^{-1} \Phi_N^T C^{-1} Y_N$$

Advantage: It is B.L.U.E.

Disadvantage: Alternate estimation of system and the colour C is necessary.