

Parameterization of Information matrix for MIMO systems with input process with finite dimensional spectrum parameterization

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Consider the parameterized model

$$y_t = G(q, \theta)u_t + H(q, \theta)e_t, \quad (1)$$

where $\theta \in \mathbf{R}^n$, $u_t \in \mathbf{R}^m$, $y_t \in \mathbf{R}^p$ and $e_t \in \mathbf{R}^p$. The input is filtered Gaussian process while the noise is a white, Gaussian process with covariance matrix Λ . The inverse per-sample covariance matrix of the parameter estimates is

$$\mathcal{I} = \frac{1}{2\pi} \int_{-\pi}^{\pi} \Gamma(e^{i\omega}) (\Lambda^{-1} \otimes \phi_{\chi_o}) \Gamma^*(e^{i\omega}) d\omega, \quad (2)$$

where

$$\phi_{\chi_o}(\omega) \triangleq \begin{bmatrix} \phi_u(\omega) & \phi_{ue}(\omega) \\ \phi_{eu}(\omega) & \Lambda \end{bmatrix}, \quad \Gamma(e^{i\omega}) \triangleq \begin{bmatrix} (\text{vec } \mathbf{F}_1^T)^T \\ \vdots \\ (\text{vec } \mathbf{F}_n^T)^T \end{bmatrix}, \quad \mathbf{F}_i \triangleq \begin{bmatrix} H^{-1}(q, \theta) \frac{\partial G}{\partial \theta_i} & H^{-1}(q, \theta) \frac{\partial H}{\partial \theta_i} \end{bmatrix}.$$

The spectrum is parameterized using a finite-dimensional parameterization given by

$$\phi_{\chi_o}(\omega) = \sum_{k=-M}^M C_k \mathcal{B}_k(e^{i\omega}), \quad (3)$$

where $\mathcal{B}_{-k}(z) = \mathcal{B}_k(z^{-k})$, $C_k \in \mathbf{R}^m$ and $C_{-k} = C_k^T$.

Vectorizing (3) gives

$$\begin{aligned} \text{vec } \mathcal{I} &= \frac{1}{2\pi} \int_{-\pi}^{\pi} \left[(\Gamma^*(e^{i\omega}))^T \otimes \Gamma(e^{i\omega}) \right] \text{vec} (\Lambda^{-1} \otimes \phi_{\chi_o}(\omega)) d\omega \\ &= \frac{1}{2\pi} \sum_{k=-M}^M \int_{-\pi}^{\pi} \left[\Gamma(e^{-i\omega}) \otimes \Gamma(e^{i\omega}) \right] \mathcal{B}_k(e^{i\omega}) d\omega \text{vec} (\Lambda^{-1} \otimes C_k) \\ &= [R_{-M} \quad \cdots \quad R_{-1} \quad R_0 \quad R_1 \quad \cdots \quad R_M] \begin{bmatrix} \text{vec} (\Lambda^{-1} \otimes C_M^T) \\ \vdots \\ \text{vec} (\Lambda^{-1} \otimes C_1^T) \\ \text{vec} (\Lambda^{-1} \otimes C_0) \\ \text{vec} (\Lambda^{-1} \otimes C_1) \\ \vdots \\ \text{vec} (\Lambda^{-1} \otimes C_M) \end{bmatrix}, \end{aligned}$$

where

$$R_k \triangleq \frac{1}{2\pi} \int_{-\pi}^{\pi} \left[\Gamma(e^{-i\omega}) \otimes \Gamma(e^{i\omega}) \right] \mathcal{B}_k(e^{i\omega}) d\omega.$$

The R_k s are not covariance matrices, in fact they are in general not square. However, the elements can be obtained from the covariances of a system formed by vectorizing $\Gamma(e^{i\omega})$. First, let's consider the above Kronecker product, i.e.,

$$(\Gamma^*(e^{i\omega}))^T \otimes \Gamma(e^{i\omega}) = \begin{bmatrix} \Gamma_{1,1}(e^{-i\omega})\Gamma(e^{i\omega}) & \Gamma_{1,2}(e^{-i\omega})\Gamma(e^{i\omega}) & \cdots & \Gamma_{1,p+m}(e^{-i\omega})\Gamma(e^{i\omega}) \\ \Gamma_{2,1}(e^{-i\omega})\Gamma(e^{i\omega}) & \Gamma_{2,2}(e^{-i\omega})\Gamma(e^{i\omega}) & \cdots & \Gamma_{2,p+m}(e^{-i\omega})\Gamma(e^{i\omega}) \\ \vdots & \vdots & \ddots & \vdots \\ \Gamma_{n,1}(e^{-i\omega})\Gamma(e^{i\omega}) & \Gamma_{n,2}(e^{-i\omega})\Gamma(e^{i\omega}) & \cdots & \Gamma_{n,p+m}(e^{-i\omega})\Gamma(e^{i\omega}) \end{bmatrix}. \quad (4)$$

Now consider the matrix

$$\begin{aligned} & \text{vec } \Gamma(e^{i\omega}) [\text{vec } \Gamma(e^{i\omega})]^* \\ &= \begin{bmatrix} \text{vec } (\Gamma(e^{i\omega}))\Gamma_{1,1}(e^{-i\omega}) & \text{vec } (\Gamma(e^{i\omega}))\Gamma_{2,1}(e^{-i\omega}) & \cdots & \text{vec } (\Gamma(e^{i\omega}))\Gamma_{n,1}(e^{-i\omega}) & \text{vec } (\Gamma(e^{i\omega}))\Gamma_{1,2}(e^{-i\omega}) \\ \text{vec } (\Gamma(e^{i\omega}))\Gamma_{2,2}(e^{-i\omega}) & \cdots & \text{vec } (\Gamma(e^{i\omega}))\Gamma_{n,2}(e^{-i\omega}) & \cdots & \text{vec } (\Gamma(e^{i\omega}))\Gamma_{n,p(p+m)}(e^{-i\omega}) \end{bmatrix}. \quad (5) \end{aligned}$$

The elements of (4) can be formed by suitably reshaping the columns of (5). The integrals

$$\tilde{R}_k \triangleq \frac{1}{2\pi} \int_{-\pi}^{\pi} \text{vec } \Gamma(e^{i\omega}) [\text{vec } \Gamma(e^{i\omega})]^* \mathcal{B}_k(e^{i\omega}) d\omega$$

can be evaluated using numerically robust methods. Note that $\tilde{R}_{-k} = \tilde{R}_k^*$.