

## 8.2: Second-Order Properties of Multivariate Time Series Examples

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# Motivation

We analyze vector-valued time series from a second-order point of view.

## Example 8.2.1

Let's focus on stationary time series. Recall the autocovariance function:

$$\Gamma(h) = E[(\mathbf{X}_{t+h} - \boldsymbol{\mu})(\mathbf{X}_t - \boldsymbol{\mu})']$$

$$\begin{bmatrix} X_{t,1} \\ X_{t,2} \end{bmatrix} = \begin{bmatrix} Z_t \\ Z_t \end{bmatrix} + \begin{bmatrix} 0 \\ .75Z_{t-10} \end{bmatrix}$$

- ①  $\Gamma(-10) = ?$
- ②  $\Gamma(0) = ?$
- ③  $\Gamma(10) = ?$

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$$\Gamma(0) = \begin{bmatrix} 1 & 1 \\ 1 & 1.5625 \end{bmatrix}$$

$$\Gamma(10) = \begin{bmatrix} 0 & 0 \\ .75 & .75 \end{bmatrix}$$

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$$R(-10) = \begin{bmatrix} 0 & .6 \\ 0 & .48 \end{bmatrix}$$

$$R(0) = \begin{bmatrix} 1 & .8 \\ .8 & 1 \end{bmatrix}$$

$$R(10) = \begin{bmatrix} 0 & 0 \\ .6 & .48 \end{bmatrix}$$

# The Autocovariance Function

## Basic Properties of $\Gamma$

- 1  $\Gamma(h) = \Gamma'(-h)$
- 2 for all  $i, j$ ,  $|\gamma_{ij}(h)| \leq \sqrt{\gamma_{ii}(0)\gamma_{jj}(0)}$
- 3  $\gamma_{ii}(h)$  is element  $i$ 's autocovariance function
- 4  $\sum_{j=1}^n \sum_{k=1}^n \mathbf{a}_j' \Gamma(j-k) \mathbf{a}_k \geq 0$

Try proving these.

# Vector-valued noise

## White Noise

The  $m$ -variate series  $\{\mathbf{Z}_t\}$  is **white noise with mean 0 and covariance matrix  $\Sigma$** , written

$$\{\mathbf{Z}_t\} \sim WN(\mathbf{0}, \Sigma),$$

if  $\{\mathbf{Z}_t\}$  is stationary with mean vector  $\mathbf{0}$  and covariance matrix function

$$\Gamma(h) = \begin{cases} \Sigma, & h = 0 \\ \mathbf{0}, & \text{otherwise.} \end{cases}$$

## IID Noise

The  $m$ -variate series  $\{\mathbf{Z}_t\}$  is **iid noise with mean 0 and covariance matrix  $\Sigma$** , written

$$\{\mathbf{Z}_t\} \sim iid(\mathbf{0}, \Sigma),$$

if  $\{\mathbf{Z}_t\}$  are independent and identically distributed with mean vector  $\mathbf{0}$  and covariance matrix  $\Sigma$ .



# More definitions

## linear process

The m-variate series  $\{\mathbf{X}_t\}$  is a **linear process** if it has the representation

$$\mathbf{X}_t = \sum_{j=-\infty}^{\infty} C_j \mathbf{Z}_{t-j}, \quad \{\mathbf{Z}_t\} \sim WN(\mathbf{0}, \Sigma)$$

where  $\{C_j\}$  is a sequence of matrices whose components are absolutely summable.

This is stationary and has covariance matrix function

$$\Gamma(h) = \sum_{j=-\infty}^{\infty} C_{j+h} \Sigma C_j'.$$

(This is a HW question)

## MA( $\infty$ ) process

An MA( $\infty$ ) process is a linear process with  $C_j = 0$  for  $j < 0$ .

$$\mathbf{x}_t = \sum_{j=0}^{\infty} C_j \mathbf{z}_{t-j}, \quad \{\mathbf{z}_t\} \sim WN(\mathbf{0}, \Sigma)$$