

Harmonizable VARMA time series

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Chapel Hill, March 2025

Abstract

- Harmonizable time series form a wide class of nonstationary time series,
- Tractable Fourier analysis, spectral distributions with correlated components
- Many fields of application, for example, analysis of replicated ElectroEncephaloGram signals for studying the brain connectivity.

Here

- Parametric form : Harmonizable Vector AutoRegressive and Moving Average models (HVARMA)
- Same spirit as of standard VARMA models: solution of a difference equation
- Harmonizable noise: uncorrelated, but variance not constant over time
- Give their spectral characteristics
- Provide a methodology for generating realizations of harmonizable time series based on known spectral characteristics.
- Discuss how to estimate the parameters on a simple example of an univariate HVARMA (1,0) model.
- Do not assume any model for the harmonizable noise except its heteroskedasticity.

Aim of the talk:

Review and complements on multivariate linear harmonizable models :
(Rozanov, 1959), (Rao, 1982), (Mehlman 1991).

Starting point of this review:

Construction and simulation of harmonizable time series with specific spectrum.

EEG and spectral analysis

- Spectral analysis is essential tool for neuroscientists. The spectrum of EEG time series has a direct neurophysiological interpretation.
- Multivariate time series problem: coherence analysis. But with such complex processes, the study of functional connectivity requires dual-frequency coherence.
(Neuroscience justifies the importance of capturing dependencies between different frequency bands at different locations)
- Dynamic phenomenon
- Several subjects: can we identify differences in subject specific reaction patterns ?
- ...

Spectral analysis: illustration

Consider an univariate time series $(X_n)_{n \in \mathbb{Z}}$.

We observe replicated realizations $\{X_n^r : t = 1, \dots, N; r = 1, \dots, R\}$.

Define its Loève spectrum as

$$\tilde{m}_{X,X} = \frac{1}{4\pi^2 R} \sum_{r=1}^R d_X^r(\omega_1) \overline{d_X^r(\omega_2)}, \quad \omega_1, \omega_2 \in [0, \pi]$$

where $d_X^r(\omega)$ is the Fourier transform of the r -th replicate.

- *Fig.1* : ElectroEncephaloGram
- *Fig.2* : Stationary time series
- *Fig.3* : Locally stationary process
- *Fig.4* : PC time series

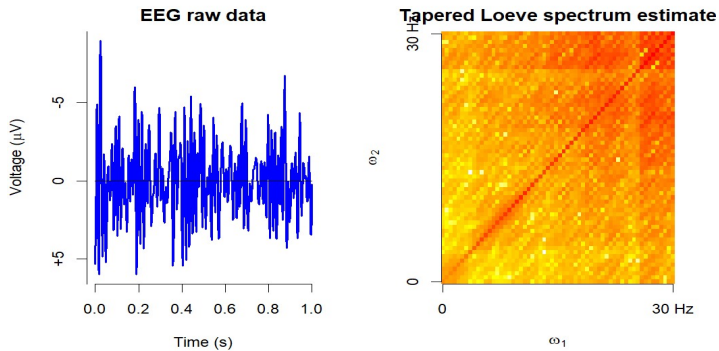


Figure: 1 ElectroEncephalGram

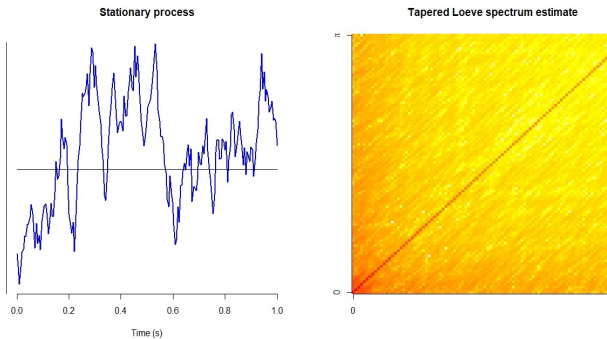


Figure: 2 Stationary time series

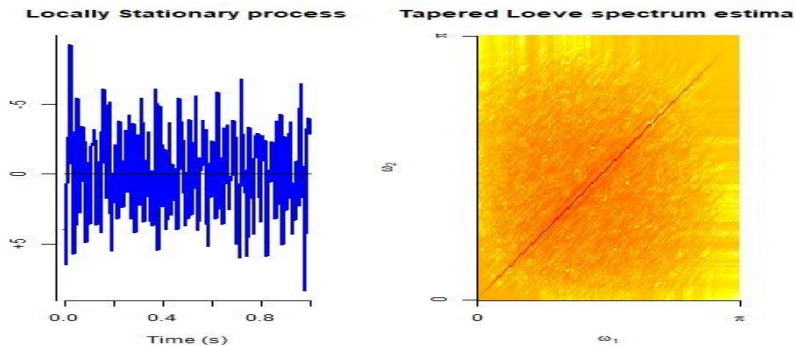


Figure: 3 Locally stationary time series

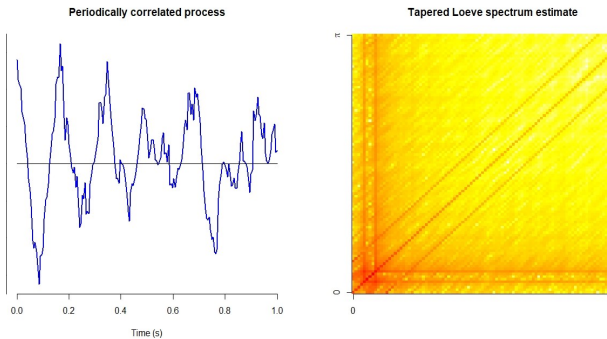


Figure: 4 PC time series

Context

- Second order structure.
- How to capture correlation across different frequencies?
That is where harmonizable processes comes in (Loève, 1945).
- Consider $(\epsilon_n)_{n \in \mathbb{Z}}$ be a real-valued white noise :

$$\mathbb{E}(\epsilon) = 0 \quad \text{and} \quad \mathbb{E}(\epsilon_n \epsilon_m) = \mathbb{I}_{\{n=m\}}.$$

Let $Z_n = \epsilon_n$ for $n \geq 0$ and $Z_n = 0$ for $n < 0$.

Then $(Z_n)_{n \in \mathbb{N}}$ is not stationary.

Question: What is the spectral structure of $(Z_n)_{n \in \mathbb{Z}}$?

Answer: Weakly harmonizable not strongly harmonizable (Abreu 1970).

Context

I – Multivariate harmonizable time series

II – Harmonizable noise

III – Harmonizable linear model

IV – Harmonizable VARMA model

V – Estimation of the parameter of an Harmonizable AR(1) model

Notation

In what follows, we solely consider centered square-integrable d -variate time series and we model their second order structure via frequencies correlation.

For $X = (X_1, \dots, X_d)^\top$ and $Y = (Y_1, \dots, Y_d)^\top$ in $\mathcal{H}^d = L^2_{\mathbb{R}^{1 \times d}}(P)$

$$\text{Cov}[X, Y] = E[XY^\top] = \left(E[X_j Y_k] \right)_{j,k=1,\dots,d} \quad \text{and} \quad \text{Var}[X] \stackrel{\text{def}}{=} \text{Cov}[X, X]$$

$\mathbb{C}^d$ denotes d -dimensional row-vector complex numbers

$\mathbb{C}^{1 \times d}$ denotes d -dimensional column vector complex numbers

$$\mathbb{T} = \mathbb{R}/2\pi \sim (-\pi, \pi]$$

Here, $^\top$ denotes the transpose operator, and $*$ denotes the adjoint operator, i.e. the conjugate transpose operator.

I – Multivariate harmonizable time series

Definition ((Rozanov 1959), (Rao 1982), (Mehlman 1991))

A d -variate time series $(X_n)_{n \in \mathbb{Z}}$ is said to be harmonizable when it admits a spectral stoch. measure $\mu_X : \mathcal{B}(\mathbb{T}) \rightarrow \mathcal{H}^d = L^2_{\mathbb{C}^{1 \times d}}(\mathbb{P})$, (σ -additive) such that

$$X_n = \int_{\mathbb{T}} e^{in\lambda} d\mu_X(\lambda).$$

The integral is in the sense of Dunford-Schwartz, 1957.

Proposition

A centered time series X is harmonizable if and only if it admits a spectral bimeasure $M_X : \mathcal{B}(\mathbb{T}) \times \mathcal{B}(\mathbb{T}) \rightarrow \mathbb{C}^{d \times d}$ (σ -additive w.r.t. each component) such that

$$\text{cov}[X_m, X_n] = \int_{\mathbb{T}} \int_{\mathbb{T}} e^{i(m\lambda_1 - n\lambda_2)} dM_X(\lambda_1, \lambda_2).$$

The above integral is a Morse-Transue integral (Rao, 1982).

- Relationship between spectral stoch. meas. and spectral bimeasure:

$$M_X(A, B) = \text{Cov}[\mu_X(A), \mu_X(B)] = E[\mu_X(A) \mu_X(B)^*].$$

- The bimeasure M_X is not necessarily extendible as a meas. on $\mathbb{T}^2$.
- When the bimeasure M_X is extendible as a meas. on $\mathbb{T}^2$, the time series is said to be strongly harmonizable, (that is, “harmonizable in the sense of Loève”.)

Otherwise it is called weakly harmonizable (or simply, harmonizable).

- The 2nd order stationary times series are strongly harmonizable.
- A centered strongly harmonizable time series X is stationary if and only if its spectral bimeasure M_X is concentrated on the first diagonal of $\mathbb{T}^2$:

$$M_X(A, B) = 0, \quad \text{for } A \cap B = \emptyset, A, B \in \mathcal{B}(\mathbb{T}).$$

That is its stoch. meas. μ_X is orthogonally scattered:

$$E[\mu_X(A)\mu_X(B)^*] = \text{Cov}[\mu_X(A), \mu_X(B)] = 0, \quad \text{when } A \cap B = \emptyset.$$

Then the spectral meas. m_X on $\mathbb{T}$ of the stationary time series X :

$$m_X(A \cap B) = M_X(A, B) = M_X(B, A)^*, \quad \text{for any } A, B \in \mathcal{B}(\mathbb{T}).$$

II – Multivariate harmonizable noise

Recall : we seek for a parametric form for harmonizable time series with an intelligible spectrum.

- Need to ensure the harmonizability of our ARMA process.
- Proceed alike to the stationary case : define a VARMA harmonizable time series as the unique solution of difference equation based on an harmonizable noise.
- Harmonizable noises to be considered as innovation for harmonizable models.

Main ingredient: "the linear space of weakly harmonizable processes is a module over the class of all bounded linear operators on $\mathcal{H}^d$ " (Rao, 1984).

Recall

Definition

A zero-mean d -variate time series $(\varepsilon_n)_{n \in \mathbb{Z}}$ is a d -variate white noise when

$$\mathbb{E}[\varepsilon_m \varepsilon_n^\top] = \text{Cov}[\varepsilon_m, \varepsilon_n] = \Sigma_\varepsilon^2$$

for some symmetrical non-negative-definite $d \times d$ -matrix Σ_ε^2 .

A d -variate white noise is a stationary time series.

Definition

In this work a d -variate pure white noise is a d -variate white noise such that

$$\Sigma_\varepsilon^2 = \mathbb{I}_{d \times d} \quad (d \times d\text{-identity matrix}).$$

Lemma (Fundamental lemma)

Let $Z_n \stackrel{\text{def}}{=} (Z_{n,1}, \dots, Z_{n,d})^\top$ such that

$$\mathbb{E}[Z_n] = 0_d \quad \text{and} \quad \mathbb{E}[Z_n Z_m^\top] = \text{Cov}[Z_n, Z_m] = \Sigma_{Z_n}^2 \mathbb{I}_{\{m=n\}}.$$

We have equivalence between:

- Condition (Ha): $\sup_{n,j} \mathbb{E}[|Z_{n,j}|^2] < \infty$.
- There exist a d -variate white noise $(\varepsilon_n)_{n \in \mathbb{Z}}$ and a family of $d \times d$ -matrices $S_n = (s_{n,j,k})_{j,k=1,\dots,d}$ such that

$$\sup_{n,j,k} |s_{n,j,k}| < \infty \quad \text{and} \quad Z_n = S_n \varepsilon_n, n \in \mathbb{Z}.$$

Then $\text{Var}[Z_n] = S_n \Sigma_\varepsilon S_n^*$.

For a pure white noise $\text{Var}[Z_n] = S_n S_n^*$, $\|\Sigma_{Z_n}^2\| = \mathbb{E}[Z_n^\top Z_n] = \text{tr}(S_n S_n^*)$.

Harmonizability of a non-correlated time series $(Z_n)_{n \in \mathbb{Z}}$

Condition **(Ha)**: $Z_n \stackrel{\text{def}}{=} (Z_{n,1}, \dots, Z_{n,d})^\top$ such that $\mathbb{E}[Z_n] = 0_d$

$$\mathbb{E}[Z_n Z_m^\top] = \text{Cov}[Z_n, Z_m] = \Sigma_{Z_n}^2 \mathbb{I}_{\{m=n\}} \quad \text{and} \quad \sup_n \|\Sigma_{Z_n}^2\| < \infty.$$

Theorem (Definition of harmonizable noise)

A time series $Z = (Z_n)_{n \in \mathbb{Z}}$ fulfilling condition (Ha) is harmonizable.

Such a time series Z is called harmonizable noise.

Its spectral stoch. meas. μ_Z defined on $\mathcal{B}(\mathbb{T})$ verifies

$$\mu_Z(A) = \frac{1}{2\pi} \sum_{n \in \mathbb{Z}} \left(\int_A e^{-in\lambda} d\lambda \right) Z_n,$$

the convergence being in $\mathcal{H}^d(Z) \subset L^2_{\mathbb{C}^{1 \times d}}(\mathbb{P})$.

Remarks

- Then the spectral bimeasure M_Z of $(Z_n)_{n \in \mathbb{Z}}$ is defined on $\mathcal{B}(\mathbb{T}) \times \mathcal{B}(\mathbb{T})$ verifies

$$M_Z(A, B) = E[\mu_Z(A)\mu_Z^*(B)] = \frac{1}{4\pi^2} \sum_{n \in \mathbb{Z}} \left(\int_A \int_B e^{-in(\lambda_1 - \lambda_2)} d\lambda_1 d\lambda_2 \right) \Sigma_{Z_n}^2.$$

- If $(\varepsilon_n)_{n \in \mathbb{Z}}$ is a white noise, then it is stationary. Hence harmonizable. The spectral stoch. meas. μ_ε is orthogonally scattered, that is,

$$M_\varepsilon(A, B) \stackrel{\text{def}}{=} E[\mu_\varepsilon(A)\mu_\varepsilon^*(B)] = 0_{d \times d} \quad \text{when} \quad A \cap B = \emptyset.$$

Moreover, it admits a spectral meas. m_ε defined on $\mathcal{B}(\mathbb{T})$:

$$m_\varepsilon(A \cap B) = M_\varepsilon(A, B) = E[\mu_\varepsilon(A)\mu_\varepsilon^*(B)], \quad A, B \in \mathcal{B}(\mathbb{T}).$$

- Let Z be a d -variate harmonizable noise.

Then every function $f \in L^2(\mathbb{T})$ is integrable w.r.t. μ_Z and

$$\left\| \int_{\mathbb{T}} f(\lambda) d\mu_Z(\lambda) \right\|_{\mathcal{H}^d}^2 \leq \frac{\varsigma^2 d}{2\pi} \int_{\mathbb{T}} |f(\lambda)|^2 d\lambda$$

where $\varsigma^2 = \sup_{n,j,k} |s_{n,j,k}|^2$.

Thus the spectral stoch. meas. μ_Z is absolutely continuous w.r.t. Lebesgue meas. on $\mathbb{T}$.

Nevertheless, this does not mean that there exists a an integrable function $\Delta_Z : \mathbb{T} \rightarrow \mathcal{H}^d = L^2_{\mathbb{C}^{1 \times d}}(\mathbb{P})$ such that

$$\mu_Z(A) = \int_A \Delta_Z(\lambda) d\lambda.$$

Nor that the spectral bimeasure M_Z is extendible to a $\mathbb{C}^{d \times d}$ -valued meas. on $\mathcal{B}(\mathbb{T}^2)$.

Examples of harmonizable noises

Let $\varepsilon \stackrel{\text{def}}{=} (\varepsilon_n)_{n \in \mathbb{Z}}$ a d -variate white noise. The following time series are harmonizable noises.

- $Z_n^{(0)} \stackrel{\text{def}}{=} \varepsilon_n$ for any $n \in \mathbb{Z}$.
- $Z_n^{(1)} \stackrel{\text{def}}{=} \varepsilon_n$ for $n \geq 0$, and $Z_n^{(1)} \stackrel{\text{def}}{=} 0$ otherwise (Abreu, 1970)
- $Z_n^{(2)} \stackrel{\text{def}}{=} S_n^{(4)} \varepsilon_n$, where $d = 2$ and $S_n^{(4)} \stackrel{\text{def}}{=} \begin{pmatrix} s_{1,1,n} & s_{1,2,n} \\ s_{2,1,n} & s_{2,2,n} \end{pmatrix}$, with $\sup_{i,j,n} |s_{i,j,n}| < \infty$.

Examples (seq.)

- Let $X \stackrel{\text{def}}{=} (X_n)_{n \in \mathbb{Z}}$ be $L^2_{\mathbb{C}^{1 \times d}}(\mathbb{P})$ -bounded time series.

Consider the *innovation series* $Z^{(3)} = (Z_n^{(3)})_{n \in \mathbb{Z}}$ defined by

$$Z_{n,j}^{(3)} \stackrel{\text{def}}{=} X_{n,j} - P_{n-1}X_{n,j},$$

where $P_{n-1}X_{n,j}$ is the one-step prediction of $X_{n,j}$, that is,

the orthogonal projection of $X_{n,j}$ onto the sub-Hilbert space $\mathcal{H}_{n-1}(X)$ of $L^2_{\mathbb{C}}(\mathbb{P})$ generated by $\{X_{p,k} : p \leq n-1, k = 1, \dots, d\}$.

Then the series $Z^{(i)} = (Z_n^{(i)})_{n \in \mathbb{Z}}$, $i = 0, \dots, 3$, are harmonizable noises.

Theorem (Strongly harmonizable noise)

A d -variate harmonizable noise $(Z_n)_{n \in \mathbb{Z}}$ is strongly harmonizable if and only if

$$\Sigma_{Z_n}^2 = \int_{\mathbb{T}} e^{in\lambda} d\mathbf{m}_Z(\lambda)$$

where $\mathbf{m}_Z$ is some $\mathbb{C}^{d \times d}$ -valued meas. on $\mathbb{T}$.

- A white noise $(\varepsilon_n)_{n \in \mathbb{Z}}$ is strongly harmonizable and $\mathbf{m}_\varepsilon(A) = \Sigma_\varepsilon \delta_0(A)$ where δ_0 is Dirac meas. at 0.

- Let $G \in L^1_{\mathbb{C}^{d \times d}}(\mathbb{T})$, the ("convolution") function

$$D(\lambda) \stackrel{\text{def}}{=} \int_{\mathbb{T}} G(u)G(u + \lambda)^* du$$

has positive semi-definite Fourier matrix-coefficients. Thus there is a strongly harmonizable noise Z such that

$$\Sigma_{Z_n}^2 = \int_{\mathbb{T}} e^{in\lambda} D(\lambda) d\lambda$$

Existence of a spectral stoch. density

Existence of a strongly measurable Lebesgue-integrable function $\Delta_Z : \mathbb{T} \rightarrow \mathcal{H}^d(Z)$ such that

$$Z_n = \int_{\mathbb{T}} e^{in\lambda} \Delta_Z(\lambda) d\lambda.$$

Theorem (stoch. density of noise)

The stoch. meas. μ_Z admits a density w.r.t. Lebesgue meas. on $\mathbb{T}$ if and only if the total energy of Z is finite :

$$\sum_n \text{tr} \Sigma_{Z_n}^2 = \sum_n \text{tr} \text{Var}[Z_n] = \sum_n \mathbb{E}[Z_n^\top \overline{Z_n}] = \sum_n \sum_j \mathbb{E}(Z_{n,j}^2) < \infty.$$

In this case, the time series Z is strongly harmonizable: the spectral bimeasure M_Z is extensible to a complex meas. on $\mathbb{R}^2$ with density

$$F_Z(\lambda_1, \lambda_2) = D_Z(\lambda_1 - \lambda_2) \stackrel{\text{def}}{=} \frac{1}{4\pi^2} \sum_{n \in \mathbb{Z}} e^{-in(\lambda_1 - \lambda_2)} \Sigma_{Z_n}^2.$$

$$\Delta_Z(\lambda) = \frac{1}{2\pi} \sum_{n \in \mathbb{Z}} e^{-in\lambda} Z_n, \quad \text{the convergence being in } L_{\mathbb{C}^d}^2(P).$$

Existence of a spectral stoch. density (seq.)

In this case, the spectral bimeasure M_Z is extendible to a $\mathbb{C}^{d \times d}$ -valued meas. on $\mathbb{T}^2$ with density

$$E[\Delta_Z(\lambda_1)\Delta_Z^*(\lambda_2)] = D_Z(\lambda_1 - \lambda_2)$$

w.r.t. Lebesgue meas. on $\mathbb{T}^2$, and where

$$D_Z(\lambda) = \frac{1}{4\pi^2} \sum_{n \in \mathbb{Z}} e^{-in\lambda} \Sigma_{Z_n}^2.$$

The spectral stoch. density Δ_Z is stationnary w.r.t. the frequencies. Furthermore

$$\Sigma_{Z_n}^2 = \int_{\mathbb{T}} e^{i\lambda n} d(m_Z \lambda) = \int_{\mathbb{T}} e^{i\lambda n} D_Z(\lambda) d\lambda.$$

Notice that a (non-null) white noise $(\varepsilon_n)_{n \in \mathbb{Z}}$ has no spectral stoch. density : Indeed, condition $\sum_n \text{tr} \Sigma_{\varepsilon_n}^2 < \infty$ is not fulfilled since $\Sigma_{\varepsilon_n}^2 = \text{constante}$.

III – Multivariate harmonizable linear model

Stability by application of a finite time-invariant linear filter on an harmonizable time series.

Theorem (Harmonizable linear model)

Let $Z = (Z_n)_{n \in \mathbb{Z}}$ be a d -variate harmonizable noise and $(A_n)_{n \in \mathbb{Z}}$ be a family of complex $d \times d$ -matrices with $\sum_n \text{tr}(A_n A_n^*) < \infty$. Then the d -variate time series $X = (X_n)_{n \in \mathbb{Z}}$

$$X_n = \sum_{m \in \mathbb{Z}} A_{n-m} Z_m = \sum_{m \in \mathbb{Z}} A_m Z_{n-m}, \quad (\text{convergence in } \mathcal{H}(Z) \subset L^2_{\mathbb{C}^d}(\mathbb{P}))$$

is well defined and harmonizable.

Moreover,

If Z is strongly harmonizable then X is strongly harmonizable.

If Z has a spectral stoch. density then X has a spectral stoch. density.

If Z has a discrete spectral stoch. meas. then X has a discrete spectral stoch. meas..

Theorem (Harmonizable linear model, seq.)

Moreover the Fourier transform $\hat{A} : \mathbb{T} \rightarrow \mathbb{C}^{d \times d}$ is μ_Z -integrable and the spectral stoch. meas. μ_X verifies for $A \in \mathcal{B}(\mathbb{T})$

$$\mu_X(A) = \frac{1}{2\pi} \sum_{n \in \mathbb{Z}} \left(\int_A e^{-in\lambda} \hat{A}(-\lambda) d\lambda \right) Z_n = \int_A \hat{A}(-\lambda) d\mu_Z(\lambda).$$

Thus

$$X_n = \int_{\mathbb{T}} e^{in\lambda} \hat{A}(-\lambda) d\mu_Z(\lambda),$$

The spectral bimeasure of X can be expressed by

$$M_X(A, B) = \frac{1}{4\pi^2} \sum_{n \in \mathbb{Z}} \left(\int_A e^{-in\lambda} \hat{A}(-\lambda) d\lambda \right) \Sigma_{Z_n}^2 \left(\int_B e^{in\lambda} \hat{A}^*(-\lambda) d\lambda \right).$$

Remarks

(1) If Z is strongly harmonizable then X is strongly harmonizable and

$$\text{Cov}[X_n, X_m] = \frac{1}{2\pi} \iint_{\mathbb{T}^2} e^{in((u+\lambda)-m\lambda)} \hat{A}(-\lambda + u) \hat{A}^*(-\lambda) d\lambda dm_Z(u)$$

The spectral bimeasure M_X is extendible to a meas. on $\mathbb{T}^2$ and verifies

$$M_X(A, B) = \frac{1}{2\pi} \int_B \left(\int_{A-\lambda_2} \hat{A}(\lambda_1 - \lambda_2) \hat{A}^*(-\lambda_2) dm_Z(\lambda_1) \right) d\lambda_2.$$

(2) If Z has a spectral stoch. density then X has a spectral stoch. density $\Delta_X(\lambda) = \hat{A}(-\lambda)\Delta_Z(\lambda)$:

$$X_n = \int_{\mathbb{T}} e^{in\lambda} \hat{A}(-\lambda) \Delta_Z(\lambda) d\lambda.$$

Thus

$$\begin{aligned} \text{Cov}\{X_n, X_m\} &= \int_{\mathbb{T}} \left(\sum_{n \in \mathbb{Z}} e^{ij\lambda} A_{n-j} D_Z(\lambda) A_{m-j}^* \right) d\lambda \\ &= \iint_{\mathbb{T}^2} e^{-i(n\lambda_1 - m\lambda_2)} \hat{A}(-\lambda_1) D_Z(\lambda_1 - \lambda_2) \hat{A}^*(-\lambda_2) d\lambda_1 d\lambda_2. \end{aligned}$$

Recall

$$\Delta_Z(\lambda) = \frac{1}{2\pi} \sum_{n \in \mathbb{Z}} e^{-in\lambda} Z_n, \quad D_Z(\lambda) = \frac{1}{4\pi^2} \sum_{n \in \mathbb{Z}} e^{-in\lambda} \Sigma_{Z_n}^2, \quad E[\Delta_Z(\lambda_1) \Delta_Z^*(\lambda_2)] = D_Z(\lambda_1 - \lambda_2)$$

IV – Harmonizable VARMA model

A difference equation on the harmonizable noise Z

$$X_n + \phi_1 X_{n-1} + \cdots + \phi_p X_{n-p} = Z_n + \theta_1 Z_{n-1} + \cdots + \theta_q Z_{n-q}. \quad (1)$$

where $\phi_1, \dots, \phi_p, \theta_1, \dots, \theta_q \in \mathbb{C}^{d \times d}$. Let

$$\Phi(z) \stackrel{\text{def}}{=} \mathbb{I}_{d \times d} + \phi_1 z + \cdots + \phi_p z^p \quad \text{and} \quad \Theta(z) \stackrel{\text{def}}{=} \mathbb{I}_{d \times d} + \theta_1 z + \cdots + \theta_q z^q.$$

We assume that $\det \Phi(e^{i\lambda}) \neq 0$ for any λ .

Then equat. (1) admits a unique harmonizable solution :

$$X_n = \int_{\mathbb{T}} e^{in\lambda} \Phi(e^{-i\lambda})^{-1} \Theta(e^{-i\lambda}) d\mu_Z(\lambda)$$

The spectral stoch. meas. μ_X verifies

$$\mu_X(A) = \frac{1}{2\pi} \sum_{n \in \mathbb{Z}} \left(\int_A e^{-in\lambda} \Phi(e^{-i\lambda})^{-1} \Theta(e^{-i\lambda}) d\lambda \right) Z_n.$$

- If Z is a strongly harmonizable noise then X is strongly harmonizable.
- If the harmonizable noise Z is of finite energy, then Z and X has spectral stoch. densities and Loève densities F_Z and F_X w.r.t. Lebesgue meas. on $\mathbb{T}^2$

$$\text{Cov}[X_n, X_m] = \frac{1}{2\pi} \iint_{\mathbb{T}^2} e^{i(n\lambda_1 - m\lambda_2)} F_X(\lambda_1, \lambda_2) d\lambda_1 d\lambda_2$$

where

$$F_Z(\lambda_1, \lambda_2) = D_Z(\lambda_1 - \lambda_2), \quad D_Z(\lambda) = \frac{1}{4\pi^2} \sum_{n \in \mathbb{Z}} e^{-in\lambda} \Sigma_{Z_n}^2$$

and

$$F_X(\lambda_1, \lambda_2) = \Phi(e^{-i\lambda_1})^{-1} \Theta(e^{-i\lambda_1}) D_Z(\lambda_1 - \lambda_2) \Theta^*(e^{-i\lambda_2}) \Phi^*(e^{-i\lambda_2})^{-1}.$$

Remarks

- Causality (Brockwell & Davis, Theorem 3.1.1) :

$$\det \Phi(z) \neq 0 \quad \text{for } |z| \leq 1 \quad \implies \quad X_n = \sum_{j=-\infty}^n A_{n-j} Z_j = \sum_{j=0}^{\infty} A_j Z_{n-j}.$$

- Invertibility (Brockwell & Davis, Theorem 3.1.2) :

$$\det \Theta(z) \neq 0 \quad \text{for } |z| \leq 1 \quad \implies \quad Z_n = \sum_{j=-\infty}^n \Pi_{n-j} X_j = \sum_{j=0}^{\infty} \Pi_j X_{n-j}.$$

Here $\Theta(z)^{-1} \Phi(z) = \sum_{j \in \mathbb{Z}} \Pi_j z^j$

- In the two previous relations, we have the equivalence whenever the polynomial functions $\det \Phi(\cdot)$ and $\det \Theta(\cdot)$ have no common zeroes.

Example : $d = 1$ Let $Z = (Z_n)_{n \in \mathbb{Z}}$, be a univariate harmonizable noise. Consider the difference equation

$$X_n = \phi X_{n-1} + Z_n + \theta Z_{n-1}.$$

- If $|\phi| \neq 1$ Then there exists a unique harmonizable solution

$$X_n = \int_{\mathbb{T}} e^{in\lambda} \frac{1 + \theta e^{-i\lambda}}{1 - \phi e^{-i\lambda}} d\mu_Z(\lambda).$$

- If $\theta = 0$ and $0 < |\phi| < 1$ then

$$X_n = \sum_{j=-\infty}^n \phi^{n-j} Z_j = \sum_{j=0}^{\infty} \phi^j Z_{n-j}$$

- If $\theta = 0$ and $|\phi| > 1$ then

$$X_n = \sum_{j=1}^{\infty} (-1)^j \phi^{-j} Z_{n+j}.$$

Construction of an harmonizable time series

Let $N > 0$ be fixed and $\omega_j \stackrel{\text{def}}{=} \frac{2\pi j}{2N+1}$, $j = -N, \dots, N$.

Let

$$Z_n = \begin{cases} \sum_{j=-N}^N e^{in\omega_j} \mu_j & \text{for } n = -N, \dots, N \\ 0 & \text{otherwise} \end{cases}$$

where $(\mu_j)_{j \in \mathbb{Z}}$ is a periodic 2nd order stationary sequence on $\mathbb{Z}$.

Let $K_\mu(r) \stackrel{\text{def}}{=} \text{Cov}[\mu_{j+r}, \mu_j]$ for any $j, r \in \mathbb{Z}$.

Then $\mu_{j+2N+1} = \mu_j$ and $K_\mu(r + 2N + 1) = K_\mu(r)$.

Then $Z = (Z_n)_{n \in \mathbb{Z}}$ is an harmonizable noise with a spectral stoch. density.

The density of the spectral bimeasure M_Z is equal to

$$F_Z(\lambda_1, \lambda_2) = D_Z(\lambda_1 - \lambda_2), \quad D_Z(\lambda) = \frac{2N+1}{4\pi^2} \sum_{j=-N}^N W_N(\omega_j - \lambda) K_\mu(j)$$

where $W_N(\lambda) \stackrel{\text{def}}{=} \sum_{n=-N}^N e^{in\lambda} = \frac{\sin((2N+1)\frac{\lambda}{2})}{\sin(\frac{\lambda}{2})}$ Dirichlet kernel .

Actually the random variables μ_j forms a d -variate mean-zero stationary time series on $\mathbb{Z}_N = \mathbb{Z}/(2N+1) \sim \{-N, \dots, N\}$

Construction (seq.)

Then the harmonizable VARMA time series X defined by

$$X_n + \phi_1 X_{n-1} + \cdots + \phi_p X_{n-p} = Z_n + \theta_1 Z_{n-1} + \cdots + \theta_q Z_{n-q}.$$

is strongly harmonizable with bispectre

$$M_X(d\lambda_1, d\lambda_2) = F_X(\lambda_1, \lambda_2) d\lambda_1 d\lambda_2$$

$$\begin{aligned} F_X(\lambda_1, \lambda_2) &= \Phi(e^{-i\lambda_1})^{-1} \Theta(e^{-i\lambda_1}) D_Z(\lambda_1 - \lambda_2) \Theta^*(e^{-i\lambda_2}) \Phi^*(e^{-i\lambda_2})^{-1} \\ &= \frac{2N+1}{4\pi^2} \sum_{j=-N}^N W_N(\omega_j - \lambda_1 + \lambda_2) \Phi(e^{-i\lambda_1})^{-1} \Theta(e^{-i\lambda_1}) K_\mu(j) \Theta^*(e^{-i\lambda_2}) \Phi^*(e^{-i\lambda_2})^{-1}. \end{aligned}$$

Spectral representation

$$\begin{aligned} X_n &= \frac{1}{2\pi} \sum_{j=-N}^N \left(\int_{\mathbb{T}} e^{i\lambda n} \Phi(e^{-i\lambda})^{-1} \Theta(e^{-i\lambda}) W_N(\omega_j - \lambda) d\lambda \right) \mu_j \\ &= \int_{\mathbb{T}} e^{i\lambda n} \Phi(e^{-i\lambda})^{-1} \Theta(e^{-i\lambda}) \left(\frac{1}{2\pi} \sum_{j=-N}^N W_N(\omega_j - \lambda) \mu_j \right) d\lambda. \end{aligned}$$

Spectral analysis: illustration 2

We observe R replications of realizations of two time series X and Y .

Loève cross-spectrum $\tilde{m}_{X,Y}$ and dual-frequency coherence $\tilde{\rho}_{X,Y}$

$$\tilde{m}_{X,Y}(\omega_1, \omega_2) = \frac{1}{4\pi^2 R} \sum_{r=1}^R d_X^r(\omega_1) \overline{d_X^r(\omega_2)}$$

$$\tilde{\rho}_{X,Y}(\omega_1, \omega_2) = \frac{|\tilde{m}_{X,Y}(\omega_1, \omega_2)|^2}{\tilde{m}_{X,X}(\omega_1, \omega_1) \tilde{m}_{Y,Y}(\omega_2, \omega_2)}, \quad \omega_1, \omega_2 \in [0, \pi]$$

where $d_X^r(\omega)$ is the Fourier transform of the r -th replicate.

- *Fig.5* : VARMA versus HVARMA (realizations)
- *Fig.6* : HVARMA(1,1)-(2,82) (Loève cross-spect. and Dual-freq. coh.)
- *Fig.7* : VARMA versus HVARMA (realizations)
- *Fig.8* : HVARMA(3,2)-(2,82) (Loève cross-spect. and Dual-freq. coh.)

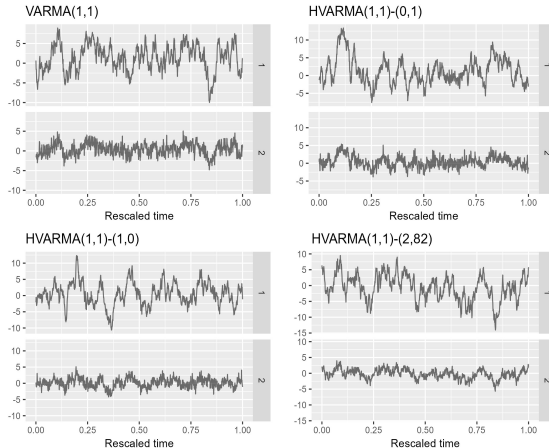


Figure: 5 VARMA versus HVARMA

HVARMA(1,1)-(2,82) and
its spectral characteristics

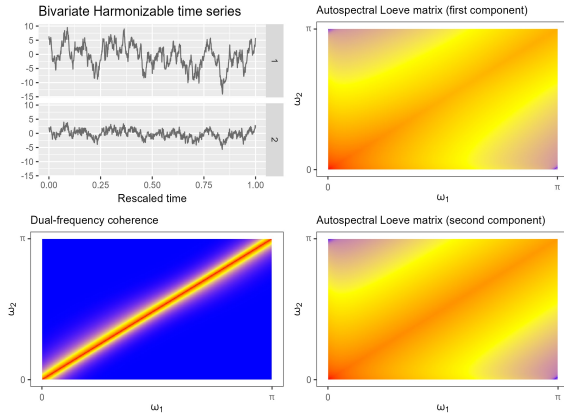


Figure: 6 HVARMA(1,1)(2,82)

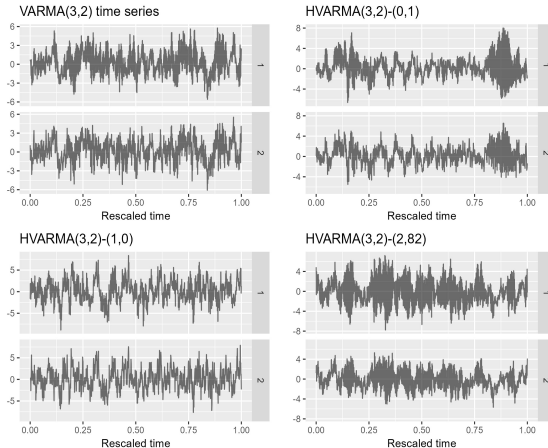


Figure: 7 VARMA versus HVARMA

HVARMA(3,2)-(2,82) and
its spectral characteristics

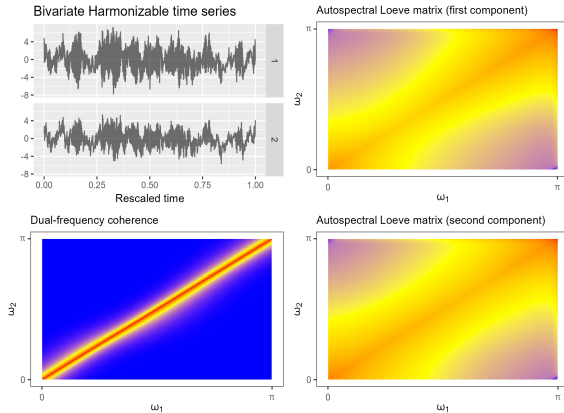


Figure: 8 HVARMA(3,2)(2,82)

VI – Parameter estimation, harmonizable AR(1), $d = 1$

H.B. Mann and A. Wald (1943), T.W. Anderson (1959), J.S. White (1958)

$$X_n - \phi X_{n-1} = Z_n.$$

X_0 and Z_n , $n \geq 1$, are square-integrable, $E(Z_n) = E(X_0) = 0$, and

$$Z_n \perp\!\!\!\perp Z_k, \quad Z_n \perp\!\!\!\perp X_0 \quad \text{for } n \neq k,$$

where $\perp\!\!\!\perp$ denotes the independence of the random variables.

Then

$$S_n \stackrel{\text{def}}{=} \sum_{k=1}^n \phi^{-k} Z_k = (\phi^{-n} X_n - X_0) \perp\!\!\!\perp X_0, Z_k, k \leq n-1$$

and

$$\text{var}(X_n) = E(X_n^2) = \phi^{2n} \sum_{k=1}^n \phi^{-2k} E(Z_k^2) + \phi^{2n} E(X_0^2).$$

Notice that, if $|\phi| > 1$ and $\sup_n E(Z_n^2) < \infty$, then $\lim_{n \rightarrow \infty} S_n = S_\infty$ and

$$\lim_{n \rightarrow \infty} \phi^{-n} X_n = S_\infty + X_0 \quad \text{in q.m. and a.e.}$$

where $E(S_\infty) = 0$ and $E(S_\infty^2) = \sum_n \phi^{-2n} \sigma_n^2 < \infty$.

Now we define the least squares estimator of $\phi \in \mathbb{R}$ as follows

$$\hat{\phi}_n \stackrel{\text{def}}{=} \arg \min_{\phi} \sum_{k=1}^n (X_k - \phi X_{k-1})^2 = \frac{\sum_{k=1}^n X_{k-1} Z_k}{\sum_{k=1}^n X_{k-1}^2}.$$

Proposition (Parameter estimation: consistency)

Let $\sup_n E(Z_n^2) < \infty$, $\sup_n E(Z_n^4) < \infty$ and $E(X_0^4) < \infty$. Assume that

(i) when $|\phi| \leq 1$, $\liminf_n \frac{1}{n} \sum_{k=1}^n E(Z_k^2) > 0$,

(ii) when $|\phi| > 1$, $P(S_\infty + X_0 = 0) = 0$.

Then

$$\lim_{n \rightarrow \infty} \hat{\phi}_n = \phi \text{ a.e.}$$

$$\text{Recall for } |\phi| > 1, \quad S_\infty \stackrel{\text{def}}{=} \sum_{k=1}^{\infty} \phi^{-k} Z_k \quad \text{q.m. and a.e.}$$

Proposition (Parameter estimation: $|\phi| < 1$ - Asymptotic normality)

Let $|\phi| < 1$, $\sup_n E(Z_n^2) < \infty$, $\sup_n E(Z_n^4) < \infty$, $E(X_0^4) < \infty$. In addition, assume that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=1}^n E(Z_j^2) = \sigma^2 > 0 \quad \text{and}$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=1}^{n-k} E(Z_j^2) E(Z_{j+k}^2) = \eta_k^4 > 0 \quad \text{for any } k \geq 1.$$

Then

$$\sqrt{n}(\hat{\phi}_n - \phi) \xrightarrow{\text{in law}} \mathcal{N}(0, \Gamma_\phi)$$

where

$$\Gamma_\phi = (1 - \phi^2)^2 \sigma^{-4} \sum_k \phi^{2(k-1)} \eta_k^4.$$

Proposition (Parameter estimation: $|\phi| = 1$ - asymptotic law)

Assume that $|\phi| = 1$, $\sup_n |\mathbb{E}(Z_n^p)| < \infty$ for any $p \geq 2$, $\mathbb{E}(X_0^4) < \infty$ and

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n \mathbb{E}(Z_k^2) = \sigma^2 > 0.$$

Then

$$n(\hat{\phi}_n - \phi) \xrightarrow{\text{in law}} \frac{B_1^2 - 1}{2 \int_0^1 B_u^2 du} \stackrel{\text{law}}{=} \frac{\int_0^1 B_u dB_u}{\int_0^1 B_u^2 du}.$$

Here $\{B_t : t \in [0, 1]\}$ be a Brownian motion.

Lemma (Parameter estimation: $|\phi| > 1$ - particular case - asymptotic law)

Let $|\phi| > 1$, $Z_n = \sigma_n \zeta_n$ with $\zeta_n \perp\!\!\!\perp X_0$, $\mathcal{L}(\zeta_n) = \mathcal{L}(\zeta)$
for some random variable ζ such $E(|\zeta|^p) < \infty$ for any $p \geq 1$ and $E(\zeta) = 0$.
Assume in addition that $\lim_{n \rightarrow \infty} \sigma_n = \varsigma < \infty$.

Then conditioned by $\{\varsigma \zeta_2 + X_0 \neq 0\}$

$$\lim_{n \rightarrow \infty} \mathcal{L}(\phi^n(\hat{\phi}_n - \phi)) = \mathcal{L}\left(\frac{(\phi^2 - 1)\varsigma \zeta_1}{\varsigma \zeta_2 + X_0}\right)$$

where $\mathcal{L}(\zeta_1) = \mathcal{L}(\zeta_2) = \mathcal{L}(\zeta)$ and ζ_1, ζ_2, X_0 are independent.

When $X_0 = 0$ and ζ Gaussian, then the law of limit is Cauchy distribution.

Proposition (Parameter estimation: $|\phi| > 1$ - Convergence in law of the mean)

If $\sup_n |\mathbb{E}(Z_n^p)| < \infty$ for any $p \geq 1$ and $\varsigma^2 \stackrel{\text{def}}{=} \lim_n n^{-1} \sum_{k=1}^n \sigma_k^2$ exists and is finite.

Then conditioned by $\{S_\infty + X_0 \neq 0\}$

$$\lim_{n \rightarrow \infty} \mathcal{L}\left(\frac{1}{\sqrt{N}} \sum_{n=1}^N \phi^n (\hat{\phi}_n - \phi)\right) = \mathcal{L}\left(\frac{(\phi^2 - 1)Z_\infty}{S_\infty + X_0}\right)$$

where the random variables Z_∞ , S_∞ and X_0 are independent with

$$\mathcal{L}(Z_\infty) = \mathcal{N}(0, (\phi - 1)^{-2} \varsigma^2) \quad \text{and} \quad S_\infty = \sum_{k=1}^{\infty} \phi^{-k} Z_k.$$

When $X_0 = 0$ and the Z_n s are Gaussian, then the law of limit is Cauchy distribution.

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THANK YOU FOR YOUR ATTENTION