

Linear Filtering, Linear Processes, ARMA Processes

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Outline

Filtering theorem

Composition and convolution

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Box-Pierce and Ljung-Box Tests

Filtering theorem

Definitions, Notations

- ▶ $l^1(\mathbb{Z})$ is the set of sequences $(\alpha_k)_{k \in \mathbb{Z}}$ with values in \mathbb{R} such that

$$\sum_{k \in \mathbb{Z}} |\alpha_k| < \infty$$

- ▶ **Linear Filtering:** Let $X = (X_t)_{t \in \mathbb{Z}}$ be a stochastic process with values in \mathbb{R} . The following sequence of variable

$$Y_t = \sum_{k \in \mathbb{Z}} \alpha_k X_{t-k}, \quad \forall t \in \mathbb{Z}$$

is called a linear filter of $(X_t)_{t \in \mathbb{Z}}$ with weights $(\alpha_k)_{k \in \mathbb{Z}}$ and we denote

$$Y = F_\alpha X$$

.

Definitions, Notations

- ▶ When the sequence $(\alpha_k)_{k \in \mathbb{Z}}$ is with finite support, i.e., $\text{card}\{k, \alpha_k \neq 0\} < \infty$, $Y = F_\alpha X$ exists and it's called a **moving-average** process associated to X .

Example: running averages

- ▶ Assume that we observe a time series (x_t)
- ▶ 7 days running average : for all $t > 7$

$$z_t = \frac{1}{7} \sum_{k=0}^6 x_{t-k} = \frac{1}{7} (x_t + x_{t-1} + \dots + x_{t-6})$$

$$\alpha_k = \begin{cases} 1/7 & \text{if } 0 \leq k \leq 6 \\ 0 & \text{Otherwise} \end{cases}$$

Theorem 1

Let $(\alpha_k)_{k \in \mathbb{Z}} \in l^1(\mathbb{Z})$ and let $(X_t)_{t \in \mathbb{Z}}$ be a stochastic process bounded in L^p for $p \geq 1$, i.e.,

$$\sup_{t \in \mathbb{Z}} \mathbb{E}(|X|^p) < \infty.$$

Let

$$\forall t \in \mathbb{Z}, \forall m, n \in \mathbb{N}, \quad Y_{t,m,n} = \sum_{k=-m}^{k=n} \alpha_k X_{t-k}$$

Then, $(Y_{t,m,n})_{m,n \geq 1}$ converge almost surely and in L^p when $(m, n) \rightarrow \infty$ to a random variable $Y_t \in L^p$, i.e.,

$$Y_{t,m,n} \rightarrow Y_t \in L^p \quad \text{and} \quad \lim_{m,n \rightarrow \infty} \mathbb{E}(|Y_{t,m,n} - Y_t|^p) = 0$$

Proof : Omitted.

Theorem 2: Filter of a stationary process

Let $(\alpha_k)_{k \in \mathbb{Z}} \in l^1(\mathbb{Z})$ and let $X = (X_t)_{t \in \mathbb{Z}}$ be a stationary stochastic process, with second order, with mean $\mu_X = \mathbb{E}(X_t)$ and ACF $t \mapsto \gamma_X(t)$. Then the filter $F_\alpha X$, defined by

$$Y_t = (F_\alpha X)_t = \sum_{k \in \mathbb{Z}} \alpha_k X_{t-k}, \quad \forall t \in \mathbb{Z},$$

exists. It's a stationary process with second order, with mean and ACF functions as follow:

$$\mu_Y = \mu_X \sum_{k \in \mathbb{Z}} \alpha_k \quad \text{and} \quad \gamma_Y(h) = \sum_{k \in \mathbb{Z}} \sum_{j \in \mathbb{Z}} \alpha_j \alpha_k \gamma_X(h+j-k), \quad \forall t \in \mathbb{Z}$$

Proof of Theorem 2

- ▶ Since X is stationary and with second order, $\|X_t^2\| = \mathbb{E}(|X_t^2|) = \gamma_X(0)^2 + |\mu_X|^2 < \infty$, then X is a bounded process. We can then apply Theorem 1 with $p = 2$ and show that the process $F_\alpha X$ exists and it belongs to L^2 .
- ▶ For the rest of the proof, we need to recall that the inner product in L^2 is continuous: if $U = \lim_{n \rightarrow +\infty} U_n$ and $V = \lim_{n \rightarrow +\infty} V_n$, then

$$\langle U, V \rangle = \lim_{n \rightarrow +\infty} \langle U_n, V_n \rangle = \left\langle \lim_{n \rightarrow +\infty} U_n, \lim_{n \rightarrow +\infty} V_n \right\rangle$$

Proof of Theorem 2

$$\begin{aligned}\mathbb{E}((F_\alpha X)_t) &= \left\langle 1, \sum_{k \in \mathbb{Z}} \alpha_k X_{t-k} \right\rangle = \left\langle 1, \lim_{n \rightarrow +\infty} \sum_{k=-n}^{k=n} \alpha_k X_{t-k} \right\rangle \\ &= \lim_{n \rightarrow +\infty} \left\langle 1, \sum_{k=-n}^{k=n} \alpha_k X_{t-k} \right\rangle \\ &= \lim_{n \rightarrow +\infty} \mathbb{E} \left(\sum_{k=-n}^{k=n} \alpha_k X_{t-k} \right)\end{aligned}$$

Proof of Theorem 2

$$\begin{aligned}\mathbb{E}((F_\alpha X)_t) &= \lim_{n \rightarrow +\infty} \sum_{k=-n}^{k=n} \alpha_k \mathbb{E}(X_{t-k}) = \lim_{n \rightarrow +\infty} \mu_X \sum_{k=-n}^{k=n} \alpha_k \\ &= \mu_X \sum_{k \in \mathbb{Z}} \alpha_k\end{aligned}$$

Exercise 1

Prove that $\gamma_Y(t) = \sum_{k \in \mathbb{Z}} \sum_{j \in \mathbb{Z}} \alpha_j \alpha_k \gamma_X(t+j-k), \quad \forall t \in \mathbb{Z}.$

Example: Filter of a WN

- ▶ Let $Z = (Z_t)_{t \in \mathbb{Z}}$ be a WN with mean 0 and variance σ^2 , i.e., $Z \sim \text{WN}(0, \sigma^2)$. Let $\alpha = (\alpha_k)_{k \in \mathbb{Z}} \in l^1(\mathbb{Z})$.
- ▶ Let $X = F_\alpha Z$ the filter associated to Z and α ,
- ▶ According to Theorem 2, $\mu_X = \mu_Z \sum_{k \in \mathbb{Z}} \alpha_k = 0$
- ▶ $\forall t \in \mathbb{Z}$

$$\gamma_X(t) = \sum_{k \in \mathbb{Z}} \sum_{j \in \mathbb{Z}} \alpha_j \alpha_k \gamma_Z(t + j - k) = \sigma^2 \sum_{j \in \mathbb{Z}} \alpha_j \alpha_{t+j}$$

- ▶ Such process is called **linear process**.

Example: ACF of a MA(q)

- ▶ Let $Z \sim \text{WN}(0, \sigma^2)$. A time series $X = (X_t)_{t \in \mathbb{Z}}$ MA(q), $q \geq 1$, is defined as follows:

$$X_t = Z_t + \sum_{k=1}^q \theta_k Z_{t-k}$$

- ▶ X is a filter of Z associated to the sequence α equal to

$$\alpha_k = \begin{cases} 1 & \text{if } k = 0 \\ \theta_k & \text{if } k = 1, \dots, q \\ 0 & \text{if } k \notin \{0, 1, \dots, q\} \end{cases}$$

Example: ACF of a MA(q)

- ▶ Using Theorem 2, $\mu_X = 0$, $\gamma_X(0) = \sigma^2 (1 + \theta_1^2 + \dots + \theta_q^2)$,
- ▶ and for $1 \leq h \leq q$, we have

$$\gamma_X(h) = \theta_h + \theta_1\theta_{h+1} + \dots + \theta_{h-q}\theta_q,$$

and for $h > q$, $\gamma_X(h) = 0$.

Exercise 2

Let (Y_t) be a time series defined by

$$Y_t = \beta t + s_t + U_t$$

where $\beta \in \mathbb{R}$, where (s_t) is periodic function with periodicity 4, and where $U = (U_t)_{t \in \mathbb{Z}}$ is a stationary process

1. Is $(Y_t)_{t \in \mathbb{Z}}$ stationary?
2. Let's define $Z = F_\alpha Y$ where $\alpha_0 = 1$, $\alpha_4 = -1$ and $\alpha_k = 0$ for all $k \neq 0$ and 4. Show that (Z_t) is a stationary process and compute γ_Z in terms of γ_U .

Backshift Operator

- ▶ The operator B defined as for

$$B^k X_t = X_{t-k}, \quad \forall t, k$$

is called the backshift operator.

- ▶ The $X = F_\alpha Z \iff$

$$X_t = P_\alpha(B)Z_t$$

where $P_\alpha(z) = \sum_{k \in \mathbb{Z}} \alpha_k z^k$

Exercise 3

Let $X = (X_t)$ be a time series defined by

$$X_t = \sum_{i=0}^k a_i t^i + Z_t, \quad \forall t \in \mathbb{Z}$$

where (Z_t) is a $\text{WN}(0, \sigma^2)$ and $a_0, a_1, \dots, a_k \in \mathbb{R}$.

1. Show that $(1 - B)X$ has a polynomial trend with degree $k - 1$
2. What happen to (X_t) when we apply the operator $(1 - B)^p$, $p \in \mathbb{N}^*$? (discuss according to p).

Causality and Invertibility

Let $Z = (Z_t)_{t \in \mathbb{Z}}$ be a stationary process, $\alpha = (\alpha_k)_{k \in \mathbb{Z}} \in l^1(\mathbb{Z})$, and let $X = F_\alpha Z$. X is called

- ▶ a **causal** process, if $\forall k < 0, \alpha_k = 0$:

$$X_t = \sum_{k=0}^{\infty} \alpha_k Z_{t-k}, \quad \text{and} \quad \sum_{k=0}^{\infty} |\alpha_k| < \infty$$

- ▶ an **invertible** process if there exists a sequence $\beta = (\beta_k)_{k \in \mathbb{Z}} \in l^1(\mathbb{Z})$ such that $\forall k < 0, \beta_k = 0$ and $Z = F_\beta X$:

$$Z_t = \sum_{k=0}^{\infty} \beta_k X_{t-k}, \quad \text{and} \quad \sum_{k=0}^{\infty} |\beta_k| < \infty$$

Causality and Invertibility, Example

Let $Z = (Z_t)_{t \in \mathbb{Z}}$ be a WN.

- **Causal** processes: $\text{MA}(q)$, $q \geq 1$:

$$X_t = Z_t + \theta_1 Z_{t-1} + \dots + \theta_q Z_{t-q}$$

- **Invertible** processes: $\text{AR}(p)$

$$X_t = \phi_1 X_{t-1} + \dots + \phi_p X_{t-p} + Z_t$$

Composition and convolution

Theorem 4: Composition of filters

Let $\alpha, \beta \in l^1(\mathbb{Z})$ and let \mathbf{X} be a stationary process, then

$$F_\alpha(F_\beta \mathbf{X}) = F_{\alpha * \beta} \mathbf{X},$$

where

$$(\alpha * \beta)_k = \sum_{j \in \mathbb{Z}} \alpha_j \beta_{k-j}$$

Composition of filters

- ▶ If $\alpha, \beta \in l^1(\mathbb{Z})$, then $\alpha * \beta \in l^1(\mathbb{Z})$
- ▶ The convolution product in $l^1(\mathbb{Z})$ is

- ▶ commutative:

$$\alpha * \beta = \beta * \alpha$$

- ▶ is distributive with respect to addition,

$$\alpha * (\beta + \gamma) = \alpha * \beta + \alpha * \gamma$$

- ▶ identity element, $e = 1_0$, $\alpha * e = \alpha$

Lemma

Let $\alpha, \beta \in l^1(\mathbb{Z})$ and the power series P_α defined on the unit circle of \mathbb{C}

$$P_\alpha(z) = \sum_{k \in \mathbb{Z}} \alpha_k z^k.$$

Then for every α and β are in $l^1(\mathbb{Z})$,

- ▶ if $P_\alpha = P_\beta$, then $\alpha = \beta$
- ▶ $P_{\alpha * \beta} = P_\alpha P_\beta$

Theorem : Invertibility of power series

Let $\alpha \in l^1(\mathbb{Z})$ such that $P_\alpha(z) = \sum_{k \in \mathbb{Z}} \alpha_k z^k$ is a polynomial. The following three statements are equivalent

1. α is invertible according to the convolution in $l^1(\mathbb{Z})$
2. If $P_\alpha(z) = 0$, then $|z| \neq 1$
3. $z \mapsto 1/P_\alpha(z)$ can be developed as a power of z in a crown in \mathbb{C} containing the unit circle, i.e., $\exists r < 1 < R$ and $\beta \in l^1(\mathbb{Z})$ such that

$$\frac{1}{P_\alpha(z)} = \sum_{k \in \mathbb{Z}} \beta_k z^k, \quad \forall r < |z| < R.$$

Theorem : Invertibility of power series (2/2)

- ▶ Where one of the previous statements is satisfied, $\alpha^{-1} = \beta$,
- ▶ β is the inverse of α
- ▶ Furthermore,

$$\forall |z| \leq 1, P_\alpha(z) \neq 0 \iff \text{Supp}(\beta) \subseteq \mathbb{N}$$

Examples (1/3)

$$\alpha = (\alpha_k), \quad \alpha_k = \begin{cases} 1 & \text{if } k = 0 \\ -1 & \text{if } k = 1 \\ 0 & \text{if } k \neq 0, 1 \end{cases}, \quad P_\alpha(z) = 1 - z$$

$$\frac{1}{P_\alpha(z)} = \frac{1}{1 - z} = \sum_{k=0}^{\infty} z^k, \quad \forall |z| < 1$$

Then

$$\alpha^{-1} = 1_{\mathbb{N}}(k) \notin l^1(\mathbb{Z}).$$

Examples (2/3)

$$\alpha = (\alpha_k), \quad \alpha_k = \begin{cases} 2 & \text{if } k = 0 \\ -1 & \text{if } k = 1 \\ 0 & \text{if } k \neq 0, 1 \end{cases}, \quad P_\alpha(z) = 2 - z$$

$$\frac{1}{P_\alpha(z)} = \frac{1}{2 - z} = \sum_{k=0}^{\infty} \left(\frac{1}{2}\right)^{k+1} z^k, \quad \forall |z| < 2$$

Then

$$\alpha^{-1} = 2^{(k+1)} 1_{\mathbb{N}}(k) \in l^1(\mathbb{Z}).$$

Examples (3/3)

$$\alpha = (\alpha_k), \quad \alpha_k = \begin{cases} 1/2 & \text{if } k = 0 \\ -1 & \text{if } k = 1 \\ 0 & \text{if } k \neq 0, 1 \end{cases}, \quad P_\alpha(z) = \frac{1}{2} - z$$

$$\frac{1}{P_\alpha(z)} = \frac{1}{\frac{1}{2} - z} = \frac{2}{1 - 2z} = \sum_{k=0}^{\infty} 2^{k+1} z^k, \quad \forall |z| < 1/2$$



Examples (3/3)

$$\begin{aligned}\frac{1}{P_\alpha(z)} &= \frac{1}{\frac{1}{2} - z} \\&= \left(-\frac{1}{z}\right) \times \frac{1}{1 - \frac{1}{2z}} \\&= \sum_{k=-\infty}^{-1} -2^{-k+1} z^k, \quad |z| > \frac{1}{2}\end{aligned}$$

Then

$$\alpha^{-1} = -2^{k+1} 1_{k \leq -1}(k) \in l^1(\mathbb{Z}).$$

Theorem 5: Identifiability

Let \mathcal{S} be the set of stationary processes with second order.

1. Let Z be a $WN(0, \sigma^2)$, for all $\alpha, \beta \in l^1(\mathbb{Z})$

$$\text{If } F_\alpha Z = F_\beta Z \Rightarrow \alpha = \beta$$

2. Let $\alpha \in l^1(\mathbb{Z})$, for all $X, Y \in \mathcal{S}$:

$$\text{If } F_\alpha X = F_\alpha Y \Rightarrow X = Y$$

Exercise 5

Prove Theorem 5

ARMA Equation

Definition (V1)

A time series $X = (X_t)_{t \in \mathbb{Z}}$ is called an **ARMA** process of order $(p, q) \in \mathbb{N} \times \mathbb{N}$ if X satisfies the following equation:

$$\forall t \in \mathbb{Z}, \quad X_t = \sum_{k=1}^p \varphi_k X_{t-k} + Z_t + \sum_{k=1}^q \theta_k Z_{t-k}$$

where

- ▶ $\varphi = (\varphi_1, \dots, \varphi_p) \in \mathbb{R}^p$,
- ▶ $\theta = (\theta_1, \dots, \theta_q) \in \mathbb{R}^q$
- ▶ $Z = (Z_t)_{t \in \mathbb{Z}}$ is a $WN(0, \sigma^2)$

$$X \sim \text{ARMA}(p, q)$$

Definition (V2)

A time series $X = (X_t)_{t \in \mathbb{Z}}$ is called an **ARMA** process of order $(p, q) \in \mathbb{N} \times \mathbb{N}$ if X satisfies the following equation:

$$\Phi(B)X = \Theta(B)Z$$

where

- ▶ B is the backshift operator
- ▶ $\Phi(z) = 1 - \varphi_1 z - \dots - \varphi_p z^p$
- ▶ $\Theta(z) = 1 + \theta_1 z + \dots + \theta_q z^q$
- ▶ $Z_t = (Z)_{t \in \mathbb{Z}}$ is a $\text{WN}(0, \sigma^2)$

Definition (V3)

A time series $X = (X_t)_{t \in \mathbb{Z}}$ is called an **ARMA** process of order $(p, q) \in \mathbb{N} \times \mathbb{N}$ if X satisfies the following equation:

$$F_{\alpha_\varphi} X = F_{\alpha_\theta} Z$$

where $([\alpha_\varphi]_k)_{k \in \mathbb{Z}}$ and $([\alpha_\theta]_k)_{k \in \mathbb{Z}}$ are sequences defined as follows

$$[\alpha_\varphi]_k = \begin{cases} 1 & \text{if } k = 0 \\ -\varphi_k & \text{if } 1 \leq k \leq p \\ 0 & \text{otherwise} \end{cases} \quad \text{and} \quad [\alpha_\theta]_k = \begin{cases} 1 & \text{if } k = 0 \\ \theta_k & \text{if } 1 \leq k \leq q \\ 0 & \text{otherwise} \end{cases}$$

AR and MA processes

A time series $X = (X)_{t \in \mathbb{Z}}$ is called

- ▶ an $AR(p)$ if X satisfies

$$\Phi(B)X = Z$$

where $\Phi(z) = 1 - \varphi_1 z - \dots - \varphi_p z^p$ and $\Theta(z) = 1$.

- ▶ a $MA(q)$ if X can be written as follows

$$X = \Theta(B)Z$$

where $\Theta(z) = 1 + \theta_1 z + \dots + \theta_q z^q$ and $\Phi(z) = 1$.

where $Z \sim \text{WN}(0, \sigma^2)$

Existence of the AR(1)

Theorem 6: AR(1) Equation: $X_t = \varphi_1 X_{t-1} + Z_t$.

- ▶ If $|\varphi_1| = 1$, there's no stationary process solution of AR(1).
- ▶ If $|\varphi_1| < 1$, AR(1) has a unique causal solution:

$$X_t = \sum_{k=0}^{\infty} \varphi_1^k Z_{t-k},$$

- ▶ If $|\varphi_1| > 1$, AR(1) has a unique non-causal solution:

$$X_t = - \sum_{k=1}^{\infty} \varphi_1^{-k} Z_{t+k} = - \sum_{k=-\infty}^{-1} \varphi_1^k Z_{t-k},$$

Proof (1/4)

- ▶ Let (X_t) be a stationary solution of second order of AR(1)
- ▶ We prove by recurrence on r , that for all $r \geq 0$

$$X_t = \sum_{k=0}^r \varphi_1^k Z_{t-k} + \varphi_1^{r+1} X_{t-r-1} \quad (*)$$

- ▶ Since $|\varphi_1| < 1$, $(\varphi_1^k)_{k \in \mathbb{Z}} \in l^1(\mathbb{Z})$ and $Z_t \in L^2$, by using Theorem 1, Chapter 2, Slide 12),

$$\sum_{k=0}^r \varphi_1^k Z_{t-k} \xrightarrow[r \rightarrow +\infty]{L^2} \sum_{k=0}^{+\infty} \varphi_1^k Z_{t-k}$$

Proof of (*) (2/4)

- ▶ $r = 0$, $X_t = \varphi_1^0 Z_t + \varphi_1^{0+1} X_{t-0-1}$
- ▶ Assume (*) is true for $r \geq 0$, according to AR(1) equation:
 $X_{t-r-1} = Z_{t-r-1} + \varphi_1 X_{t-r-2}$ Then

$$\begin{aligned} X_t &= \sum_{k=0}^r \varphi_1^k Z_{t-k} + \varphi_1^{r+1} X_{t-r-1} \\ &= \sum_{k=0}^r \varphi_1^k Z_{t-k} + \varphi_1^{r+1} (Z_{t-r-1} + \varphi_1 X_{t-r-2}) \\ &= \sum_{k=0}^{r+1} \varphi_1^k Z_{t-k} + \varphi_1^{r+2} X_{t-r-2} \end{aligned}$$

Proof (3/4)

- ▶ Since (X_t) is a stationary and second order process and $|\varphi_1| < 1$

$$\varphi_1^{r+1} X_{t-r-1} \xrightarrow[r \rightarrow +\infty]{L^2} 0.$$

- ▶ Then

$$X_t = \sum_{k=0}^{+\infty} \varphi_1^k Z_{t-k}$$

- ▶ If $|\varphi_1| > 1$, prove by recurrence that

$$X_t = - \sum_{k=1}^r \varphi_1^{-k} Z_{t+k} + \varphi_1^{-r} X_{t+r}, \quad \forall r \geq 1 \quad (**)$$

- ▶ $|\varphi_1| = 1$, AR(1)= Random Walk

Proof of (**) (4/4)

- ▶ $r = 1$, $X_t = \varphi_1^{-1}X_{t+1} - \varphi_1^{-1}Z_{t+1}$
- ▶ Assume (**) is true for $r \geq 1$, according to AR(1) equation:

$$X_{t+r} = \varphi_1^{-1}X_{t+r+1} - \varphi_1^{-1}Z_{t+r+1}$$

Then

$$\begin{aligned} X_t &= - \sum_{k=1}^r \varphi_1^{-k} Z_{t+k} + \varphi_1^{-r} X_{t+r} \\ &= - \sum_{k=1}^r \varphi_1^{-k} Z_{t+k} + \varphi_1^{-r} (-\varphi_1^{-1} Z_{t+r+1} + \varphi_1^{-1} X_{t+r+1}) \\ &= - \sum_{k=1}^{r+1} \varphi_1^{-k} Z_{t+k} + \varphi_1^{r+1} X_{t+r+1} \end{aligned}$$

Existence of the ARMA process

Theorem 7

Let Φ and Θ be the polynomials associated to the ARMA(p, q) Equation.

1. If Φ doesn't have a unit root, **then** ARMA(p, q) has a unique stationary solution, given by $F_\alpha Z$ where

$$\alpha = \alpha_\theta * \alpha_\varphi^{-1}$$

2. If the Equation ARMA(p, q) has a linear process $F_\alpha Z$ as a solution, then any unit root of Φ is also a unit root of Θ .

Proof (1)

- Since Φ has no unit root, by applying Theorem 4 (L2, Slide 35), $\exists \beta \in I^1(\mathbb{Z})$, such that

$$\beta = \alpha_\varphi^{-1}.$$

Then

$$X = F_{\beta * \alpha_\varphi} X = F_\beta (F_{\alpha_\varphi} X) = F_\beta (F_{\alpha_\theta} Z) = F_{\beta * \alpha_\theta} Z$$

Proof (2)

- ▶ $F_\alpha Z$ is a solution of ARMA(p, q), then

$$F_{\alpha_\varphi} X = F_{\alpha_\theta} Z \Rightarrow F_{\alpha_\varphi} (F_\alpha Z) = F_{\alpha_\theta} Z \Rightarrow F_{\alpha * \alpha_\varphi} Z = F_{\alpha_\theta} Z$$

- ▶ According to Theorem 5, $\alpha * \alpha_\varphi = \alpha_\theta$.

Then $P_\alpha(z)\Phi(z) = \Theta(z)$.

Causal and Invertible solution of ARMA(p, q)

Theorem 8

Let Φ and Θ be two polynomials associated to ARMA(p, q). Assume that Φ and Θ don't have common zeroes. Then

- ▶ The solution of ARMA(p, q) is causal $\iff \forall z, |z| \leq 1, \Phi(z) \neq 0$.
- ▶ The solution of ARMA(p, q) is invertible $\iff \forall z, |z| \leq 1, \Theta(z) \neq 0$.

Proof: Exercise

Example 2(1/2)

$Z \sim \text{WN}(0, \sigma^2)$, and ARMA(1, 1) Equation:

$$X_t = .5X_t + Z_t + 2Z_{t-1}$$

- ▶ $\Phi(z) = 1 - .5z$ has 2 as a root, with modulus > 1 , Then the equation above admit a unique causal solution.
- ▶ $\Theta(z) = 1 + 2z$ has $-1/2$ as a root, with modulus < 1 , Then the solution is not invertible.
- ▶ The inverse of α_φ is the sequence $\beta \in l^1(\mathbb{Z})$ are the coefficients of the power series expansion of $\frac{\Theta(z)}{\Phi(z)}$ in $|z| < 2$,

$$\frac{\Theta(z)}{\Phi(z)} = \frac{1 + 2z}{1 - .5z} = 1 + \sum_{k=1}^{\infty} 5(1/2)^k z^k$$

Example 1(2/2)

- The solution is

$$X_t = Z_t + \sum_{k=1}^{\infty} 5(1/2)^k Z_{t-k}$$

- By applying Theorem 2,

$$\gamma_X(h) = \sigma^2 \left(1_{h=0} + 5(1/2)^{|h|} 1_{h \neq 0} + (25/3)(1/2)^{|h|} \right)$$

Example 2(1/4)

- Consider the following ARMA Equation

$$X_t = .4X_{t-1} + .21X_{t-2} + Z_t + .6Z_{t-1} + .09Z_{t-2}$$

- The polynomials Φ and Θ are

$$\Phi(z) = 1 - .4z - .21z^2 = (1 - .7z)(1 + .3z),$$

$$\Theta(z) = 1 + .6z + .09z^2 = (1 + .3z)^2.$$

- The ARMA can be reset as follows

$$X_t = .7X_{t-1} + Z_t + .3Z_{t-1}.$$

Example 2(2/4)

- ▶ The roots of Φ and Θ are respectively $10/7$ and $-10/3$
- ▶ We can then compute causal and invertible solutions for the above ARMA equation.
- ▶ The solution (X_t) can be written then

$$X_t = \sum_{j=0}^{\infty} \psi_j Z_{t-j} \quad \text{and} \quad Z_t = \sum_{j=0}^{\infty} \pi_j X_{t-j}, \quad t \in \mathbb{Z},$$

Example 2(3/4)

- ▶ where, for all $|z| \leq \min(10/3, 10/7) = 10/7$

$$\sum_{j=0}^{\infty} \psi_j z^j = \psi(z) = \frac{\theta(z)}{\phi(z)} = \frac{1 + .3z}{1 - .7z} = (1 + .3z) \sum_{j=0}^{\infty} (.7z)^j,$$

and

$$\sum_{j=0}^{\infty} \pi_j z^j = \pi(z) = \frac{\phi(z)}{\theta(z)} = \frac{1 - .7z}{1 + .3z} = (1 - .7z) \sum_{j=0}^{\infty} (-.3z)^j$$

Example 2(4/4)

- ▶ Then, for all $j \in \mathbb{N}$

$$\psi_0 = 1 \quad \text{and} \quad \psi_j = (.7 + .3)(.7)^{j-1} = (.7)^{j-1},$$

and

$$\pi_0 = 1 \quad \text{and} \quad \pi_j = (-1)^j(.3 + .7)(.3)^{j-1} = (-1)^j(.3)^{j-1}.$$

- ▶ Hence,

$$X_t = Z_t + \sum_{j=1}^{\infty} (.7)^{j-1} Z_{t-j} \quad \text{and} \quad Z_t = X_t + \sum_{j=1}^{\infty} (-1)^j (.3)^{j-1} X_{t-j}.$$

Exercise

Let (Z_t) be a WN. For each of the following equations, is there a stationary process solution? If the solution exist, can you say if it's a causal or invertible solution?

1. $X_t + 0.2X_{t-1} - 0.48X_{t-2} = Z_t$;
2. $X_t + 1.9X_{t-1} + 0.88X_{t-2} = Z_t + 0.2Z_{t-1} + 0.7Z_{t-2}$;
3. $X_t + 0.6X_{t-2} = Z_t + 1.2Z_{t-1}$;
4. $X_t + 1.8X_{t-1} + 0.81X_{t-2} = Z_t$.

Solution (1/2)

1. ☒ Stationary ☒ Causal ☒ Invertible

```
> abs(polyroot(c(1,.2,-.48)))  
[1] 1.666667 1.250000
```

2. ☒ Stationary ☐ Causal ☒ Invertible

```
> abs(polyroot(c(1,1.9,.88)))  
[1] 0.9090909 1.2500000
```

```
> abs(polyroot(c(1,.2,.7)))  
[1] 1.195229 1.195229
```

Solution (2/2)

3. ☒ Stationary ☒ Causal ☐ Invertible

```
> abs(polyroot(c(1,0,.6)))
```

```
[1] 1.290994 1.290994
```

```
> abs(polyroot(c(1,1.2)))
```

```
[1] 0.8333333
```

4. ☒ Stationary ☒ Causal ☒ Invertible

```
> abs(polyroot(c(1,1.8,.81)))
```

```
[1] 1.111111 1.111111
```

Exercise

Let $X = (X_t)_{t \in \mathbb{Z}}$ be the solution of the following ARMA(2, 1) process

$$(1 - B + B^2/4)X_t = (1 + B)Z_t$$

where $Z = (Z_t)_{t \in \mathbb{Z}}$ is a $WN(0, \sigma^2)$

1. Show that the solution of the above equation is causal and can be written as follows:

$$X_t = \sum_{k \geq 0} \psi_k Z_{t-k}.$$

2. Is X invertible?
3. Compute the coefficients $(\psi_k)_{k \in \mathbb{Z}}$
4. Compute the ACF of X .
5. Verify the previous results using R

Exercise

1. What are the polynomials Φ and Θ in the following ARMA process

$$X_t - 3X_{t-1} = Z_t - \frac{10}{3}Z_{t-1} + Z_{t-2}$$

where $Z = (Z_t)_{t \in \mathbb{Z}}$ is a $WN(0, \sigma^2)$

2. Show that this ARMA Equation has a unique stationary solution.
3. Is the solution causal?
4. Compute this solution in terms of Z
5. Is the solution invertible?

Exercise

Let

- ▶ $X = (X_t)_{t \in \mathbb{Z}}$ and $Y = (Y_t)_{t \in \mathbb{Z}}$ be two centred independent processes, i.e., $\forall t, s$ Y_t and X_s are independent.
- ▶ $\epsilon = (\epsilon_t)_{t \in \mathbb{Z}}$ and $\eta = (\eta_t)_{t \in \mathbb{Z}}$ be two $WN(0, \sigma^2)$.

Assume that X and Y are ARMA(1,1) processes:

$$X_t = \varphi_1 X_{t-1} + \epsilon_t + \theta_1 \epsilon_{t-1} \text{ and } Y_t = \varphi_2 Y_{t-1} + \eta_t + \theta_2 \eta_{t-1}$$

1. Show that if $\max(|\varphi_1|, |\varphi_2|) < 1$ then X and Y are causal.
2. Show that if $\max(|\eta_1|, |\eta_2|) < 1$ then X and Y are invertible.
Assume now that $\max(|\varphi_1|, |\varphi_2|) < 1$ and $\max(|\eta_1|, |\eta_2|) < 1$.
3. Show that $\forall s, t, \text{Cov}(\epsilon_s, \eta_t) = 0$
4. Compute the ACF of $Z_t = X_t Y_t$.

Theorem, ACF of an ARMA process

Let Φ and Θ be two polynomials. Assume that

$$\forall |z| \leq 1, \Phi(z) \neq 0.$$

Let X the stationary causal process solution of the following ARMA Equation

$$\Phi(B)X = \Theta(B)Z$$

where $Z \sim \text{WN}(0, \sigma^2)$.

Then $\exists 0 \leq \rho < 1$ and $C > 0$ such that

$$\forall h \in \mathbb{Z}, |\gamma_X(h)| \leq C\rho^{|h|}$$

Proof

- ▶ X is causal, then $\exists \psi \in l^1(\mathbb{Z})$ such that $X_t = \sum_{k \in \mathbb{N}} \psi_k Z_{t-k}$
- ▶ $\psi \in l^1(\mathbb{Z})$, then $\exists 0 \leq \rho < 1$ and $C' > 0$ such that $\forall k \in \mathbb{N}$

$$|\psi_k| \leq C' \rho^k$$

- ▶ X is a linear process, then

$$\gamma_X(h) = \sigma^2 \sum_{j \in \mathbb{N}} \psi_j \psi_{j+h}$$

- ▶ Then

$$\begin{aligned} |\gamma_X(h)| &\leq \sigma^2 \sum_{j \in \mathbb{N}} (C')^2 \rho^{2j+|h|} \\ &\leq \sigma^2 \frac{(C')^2}{1 - \rho^2} \rho^{|h|} \end{aligned}$$

Box-Pierce and Ljung-Box Tests

Estimation of the Autocorrelation

- ▶ Let x_1, \dots, x_T be a realisation (observations) of a time series $X = (X_t)_{t \in \mathbb{Z}}$.
- ▶ $\forall, k = 1, \dots, T - 1$

$$\hat{\gamma}_X(k) = \frac{1}{T} \sum_{t=k+1}^T (x_t - \bar{x}_T)(x_{t-k} - \bar{x}_T),$$

and

$$\hat{\rho}_X(k) = \frac{\sum_{t=k+1}^T (x_t - \bar{x}_T)(x_{t-k} - \bar{x}_T)}{\sum_{t=1}^T (x_t - \bar{x}_T)^2}.$$

Theorem

Assume that x_1, \dots, x_T is a sequence of i.i.d random variables with second order, i.e., $\mathbb{E}(x_k^2) < \infty, \forall k = 1, \dots, T$. Then

- ▶ $\hat{\rho}_X(k), k = 1, \dots, T$ are approximately independent
- ▶ $\hat{\rho}_X(k), k = 1, \dots, T$ follow approximately a Normal distribution with mean 0 and variance $\frac{1}{T}$.

Hypothesis testing (2/2)

- ▶ Statistical test: for all $h \geq 1$

$$H_0^h : \rho_1 = \rho_2 = \dots = \rho_h = 0$$

vs

$$H_1^h : \exists k \leq h \rho_k \neq 0$$

- ▶ The test statistic:

$$Q(h) = T \sum_{j=1}^h \hat{\rho}_j^2 = \sum_{j=1}^h \left(\frac{\hat{\rho}_j - 0}{1/\sqrt{T}} \right)^2$$

Hypothesis testing (1/2)

- ▶ The test statistic (small samples):

$$Q(h) = T(T+2) \sum_{j=1}^h \frac{\hat{\rho}_j^2}{T-k}$$

- ▶ Distribution of the test statistic: under H_0^h ,

$$Q(h) \sim \chi^2(h)$$

- ▶ If the test is used on residuals obtained from a model with m parameters, then, under H_0^h ,

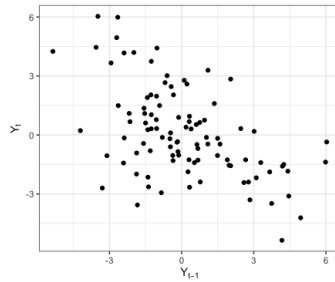
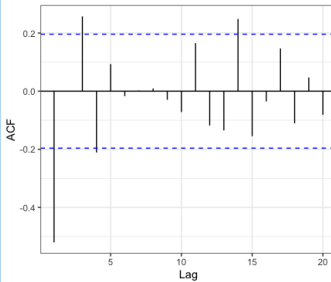
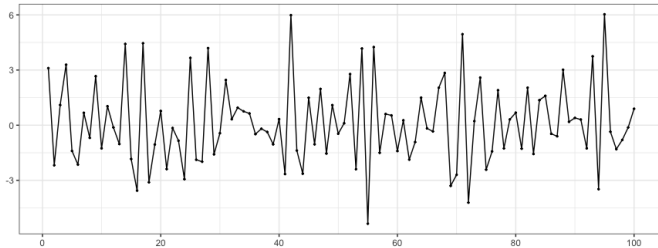
$$Q(h) \sim \chi^2(h-m)$$

Practice with R

```
> library(caschnono)
> set.seed(123)
> y1=arima.sim(n=100,list(ma=c(-.7,.3)),sd=sqrt(4))
> library(forecast)
> library(ggplot2)
> ggtsdisplay(y1, plot.type='scatter',theme = theme_bw())
```

Pract

```
> lib  
> set  
> y1=  
> lib  
> lib  
> ggt
```



bw())

Practice with R

```
> a1=Box.test.2(y1,nlag=1:10,type="Ljung-Box")
> a1
```

	Retard	p-value
[1,]	1	2.000e-07
[2,]	2	1.330e-06

```
..
> y2<-rnorm(100)
> a2=Box.test.2(y2,nlag=1:10,type="Ljung-Box")
> a2
```

	Retard	p-value
[1,]	1	0.7024042
[2,]	2	0.8247985

```
..
```

Practice with R

$$X_t = -.3X_{t-1} + .1X_{t-2} + Z_t$$

```
> library(caschnono)
> y3=arima.sim(n=100,list(ar=c(-.3,.1)),sd=sqrt(4))
> ggtsdisplay(y3, plot.type='scatter',theme = theme_bw())
> library(ZIM)
> z=y3+.3*bshift(y3,1)-.1*bshift(y3,2)
> b1=Box.test.2(y3,nlag=1:10,type="Ljung-Box")
> b1
```

	Retard	p-value
[1,]	1	0.00087823
[2,]	2	0.00046357
[3,]	3	0.00139091

```
> b2=Box.test.2(z,nlag=1:10,type="Ljung-Box")
> b2
```

	Retard	p-value
[1,]	1	0.8910890
[2,]	2	0.8204297
[3,]	3	0.7179151

Pract

```
> library(ggtsb)
> y3=as.ts(1:100)
> ggtsb(y3)
> library(ggtsb)
> z=y3-10
> b1=B
> b1
```

```
[1,]
```

```
[2,]
```

```
[3,]
```

```
> b2=B
> b2
```

```
[1,]
```

```
[2,]
```

```
[3,]
```

