

TIME SERIES ANALYSIS II

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- Partial autocorrelation function
- ARMA models for stationary time series, AR and MA models as their more simple case
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$s := v$ or $v =: s \dots$ denoting expression v by symbol s .

iff stands for *if and only if*.

Sets and mappings:

- $\mathbb{N}, \mathbb{Z}, \mathbb{R}, \mathbb{C} \dots$ natural numbers, integers, real and complex numbers, respectively.
- $\mathbb{Z}_N := \{0, 1, \dots, N - 1\} \dots$ residuals modulo $N \in \mathbb{N}$.
- $\mathbb{R}^+ \dots$ the set of all non-negative real numbers.
- $\exp X \dots$ class of all subsets of the set X .
- $\text{card } M \dots$ cardinality of a set M .
- $(\cdot)^+ : \mathbb{R} \rightarrow \mathbb{R}^+ \dots$ mapping defined as $(x)^+ = \max(0, x)$.
- $(a, b), [a, b], (a, b], [a, b) \dots$ intervals on real line.
- $J(a, b) = \{x \mid \min(a, b) < x < \max(a, b)\}$
- $J[a, b] = \{x \mid \min(a, b) \leq x \leq \max(a, b)\}$.
- $f(A) := \{y \in Y \mid y = f(x), x \in A \subseteq X\} \dots$ range (image) of set A under mapping $f : X \rightarrow Y$.
- $f^{-1}(B) := \{x \in X \mid f(x) \in B\} \subseteq X \dots$ inverse image of set $B \subseteq Y$ under mapping $f : X \rightarrow Y$.
- $I_A \dots$ indicator function of set $A \subseteq X$:

$$I_A(x) = \begin{cases} 1 & \text{for } x \in A \\ 0 & \text{otherwise} \end{cases}.$$
- $A_n \uparrow \dots$ increasing or non-decreasing sequence of numbers or sets.
- $A_n \downarrow \dots$ decreasing or non-increasing sequence of numbers or sets.
- $\sum_{i=1}^n A_i := \bigcup_{i=1}^n A_i \dots$ union of a family of sets which are pairwise disjoint.
- $A^c := X - A \dots$ complement of set $A \subseteq X$ in X where X is a priori known from the context.
- $\underline{A} := \liminf_{n \rightarrow \infty} A_n := \bigcup_{n=1}^{\infty} \bigcap_{j=n}^{\infty} A_j \dots$ inferior limit of a sequence of sets.

- $\bar{A} := \limsup_{n \rightarrow \infty} A_n := \bigcap_{n=1}^{\infty} \bigcup_{j=n}^{\infty} A_j$... superior limit of a sequence of sets.
- $A = \lim_{n \rightarrow \infty} A_n$ iff $\underline{A} = \bar{A}$, clearly
 $A_n \uparrow A$ implies $\lim_{n \rightarrow \infty} A_n = \bigcup_{n=1}^{\infty} A_n$ and
 $A_n \downarrow A$ implies $\lim_{n \rightarrow \infty} A_n = \bigcap_{n=1}^{\infty} A_n$.

Vectors and matrices:

- $\mathbf{x} := [x_1, \dots, x_n]^T$... vector of numbers (by default column vector if not stated otherwise).
- $\mathbf{x} + h := [x_1 + h, \dots, x_n + h]^T$, $h \in \mathbb{C}$.
- $\mathbf{x}_t := [x_{t_1}, \dots, x_{t_k}]^T \in \mathbb{C}^k$ where $\mathbf{t} = [t_1, \dots, t_k]^T \in \mathbb{N}^k$, $t_i \in \{1, \dots, n\}$ for $i = 1, \dots, k$.
- $\mathbf{x}(i) := [x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n]^T$ for any $1 \leq i \leq n$.
- $f(\mathbf{x}) := f(x_1, \dots, x_n)$, $d\mathbf{x} := dx_1 \dots dx_n$.
- $\mathbf{0}, \mathbf{0}_{n \times 1}$... vector of n zero entries.
- $\mathbf{A}, \mathbf{A}_{m \times n} := [a_{ij}] = [A(i, j)]$... matrix of size $m \times n$.
- $\mathcal{R}(\mathbf{A}) := \{\mathbf{y} \mid \mathbf{y} = \mathbf{A}\mathbf{x}\}$... range space of matrix operator \mathbf{A} .
- $\mathcal{N}(\mathbf{A}) := \{\mathbf{x} \mid \mathbf{A}\mathbf{x} = \mathbf{0}\}$... null space (kernel) of matrix operator \mathbf{A} .
- $\mathbf{A}^T := [a_{ji}]$... matrix transpose.
- $\mathbf{A}^* := [\bar{a}_{ji}]$... matrix adjoint.
- $\mathbf{I}, \mathbf{I}_n := \mathbf{I}_{n \times n} = [\delta_{ij}]$... identity matrix of order n .
- $\det \mathbf{A}$... determinant of a square matrix \mathbf{A} .
- $\mathbf{0}, \mathbf{0}_{m \times n}$... zero matrix of size $m \times n$.
- $\text{diag}(\mathbf{x}) := \begin{bmatrix} x_1 & 0 & \dots & 0 \\ 0 & x_2 & \dots & 0 \\ & & \vdots & \\ 0 & 0 & \dots & x_n \end{bmatrix}$... diagonal matrix.
- $\mathbf{A}(i, :)$:= $[a_{i1}, \dots, a_{in}]$... i -th row of matrix \mathbf{A} using MATLAB style.
- $\mathbf{A}(:, j)$:= $[a_{1j}, \dots, a_{mj}]^T$... j -th column of matrix \mathbf{A} using MATLAB style.

- $\mathbf{A} := [r_1; \dots; r_m] = [s_1, \dots, s_n] \dots$ forming matrix \mathbf{A} row-by-row or columnwise using MATLAB style.
- $\mathbf{A} > 0$ (or $\mathbf{A} \geq 0$) ... positively (semi)definite (non-negatively definite) matrix.
- $\langle \mathbf{x}, \mathbf{y} \rangle := \sum_{i=1}^n x_i y_i = \mathbf{y}^* \mathbf{x} \dots$ scalar (inner) product of vectors \mathbf{x} and \mathbf{y} .
- $\|\mathbf{x}\| := \sqrt{\sum_{i=1}^n |x_i|^2} = \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle} \dots$ Euclidean norm of vector \mathbf{x} .

Random variables and random vectors:

- $X \dots$ random variable.
- $\mathbb{X} := [X_1, \dots, X_n]^T \dots$ (real) random vector, indexing conventions listed above for number vectors are adopted accordingly.
- $\mu := \mu_X := EX \dots$ expectation of random variable X .
- $\boldsymbol{\mu} := \boldsymbol{\mu}_{\mathbb{X}} := E\mathbb{X} := [EX_1, \dots, EX_n]^T \dots$ expectation of random vector \mathbb{X} .
- $\sigma^2 := \sigma_X^2 := \text{var}X := E|X - EX|^2 = E|X|^2 - |EX|^2 \geq 0 \dots$ variance of random variable X .
- $\sigma_{XY} := \text{cov}(X, Y) := E(X - EX)(Y - EY) = EXY - (EX)(EY) \dots$ covariance of random variables X and Y .
- $\Sigma_{\mathbb{X}} := \text{var}\mathbb{X} := [\text{cov}(X_i, X_j)] = E(\mathbb{X} - E\mathbb{X})(\mathbb{X} - E\mathbb{X})^T = E\mathbb{X}\mathbb{X}^T - (E\mathbb{X})(E\mathbb{X})^T \dots$ variance matrix of random vector \mathbb{X} .
- $\Sigma_{\mathbb{X}\mathbb{Y}} := \text{cov}(\mathbb{X}, \mathbb{Y}) := [\text{cov}(X_i, Y_j)] = E(\mathbb{X} - E\mathbb{X})(\mathbb{Y} - E\mathbb{Y})^T = E\mathbb{X}\mathbb{Y}^T - (E\mathbb{X})(E\mathbb{Y})^T \dots$ covariance matrix of \mathbb{X} and \mathbb{Y} .

It holds:

- $\text{var}X = \text{cov}(X, X)$.
- $\text{cov}(Y, X) = \text{cov}(X, Y)$.
- $\text{cov}(\sum_r X_r, \sum_s Y_s) = \sum_r \sum_s \text{cov}(X_r, Y_s)$ and hence in particular:

- $\text{var}(X + Y) = \text{var}X + \text{cov}(X, Y) + \text{cov}(Y, X) + \text{var}Y = \text{var}X + 2\text{cov}(X, Y) + \text{var}Y$.
- $\text{cov}(\mathbb{X}, \mathbb{X}) = \text{var}\mathbb{X}$.
- $\text{cov}(\mathbb{Y}, \mathbb{X}) = \text{cov}(\mathbb{X}, \mathbb{Y})^T$ implies:
- $\text{var}\mathbb{X} = (\text{var}\mathbb{X})^T$... **variance matrix \mathbb{X} is symmetrical**.
- Given number vectors \mathbf{a} and \mathbf{c} , and matrices B and D of compatible sizes then

$$\text{cov}(\mathbf{a} + B\mathbb{X}, \mathbf{c} + D\mathbb{Y}) = \text{cov}(B\mathbb{X}, D\mathbb{Y}) = B \text{cov}(\mathbb{X}, \mathbb{Y}) D^T$$

$$\Downarrow \mathbb{X} = \mathbb{Y}$$
- $\text{var}(\mathbf{a} + B\mathbb{X}) = \text{cov}(\mathbf{a} + B\mathbb{X}, \mathbf{a} + B\mathbb{X}) = \text{cov}(B\mathbb{X}, B\mathbb{X}) = B \text{var}(\mathbb{X}) B^T$

$$\Downarrow \mathbf{a} = \mathbf{0}, B = \mathbf{b}^T$$
- $0 \leq \text{var}(\mathbf{b}^T \mathbb{X}) = \mathbf{b}^T \text{var}\mathbb{X} \mathbf{b}$ implies:
- $\text{var}\mathbb{X} \geq 0$... **variance matrix is non-negatively positive** and consequently it has non-negative eigen values λ_i and its square root matrix $\Sigma_{\mathbb{X}}^{\frac{1}{2}}$ having eigen values $\lambda_i^{\frac{1}{2}}$ may be constructed such that:
- $\Sigma_{\mathbb{X}} = \Sigma_{\mathbb{X}}^{\frac{1}{2}} \Sigma_{\mathbb{X}}^{\frac{1}{2}}$.
- $\text{cov}(\sum_r \mathbb{X}_r, \sum_s \mathbb{Y}_s) = \sum_r \sum_s \text{cov}(\mathbb{X}_r, \mathbb{Y}_s)$ and hence in particular:
- $\text{var}(\mathbb{X} + \mathbb{Y}) = \text{var}\mathbb{X} + \text{cov}(\mathbb{X}, \mathbb{Y}) + \text{cov}(\mathbb{Y}, \mathbb{X}) + \text{var}\mathbb{Y} = \text{var}\mathbb{X} + 2\text{cov}(\mathbb{X}, \mathbb{Y}) + \text{var}\mathbb{Y}$.

REFERENCES

- [1] Jiří Anděl. *Statistická analýza časových řad*. SNTL, Praha, 1976.
- [2] Peter J. Brockwell and Richard A. Davis. *Introduction to Time Series and Forecasting*. Springer-Verlag, New York, 2-nd edition, 2002.
- [3] Peter J. Brockwell and Richard A. Davis. *Time Series: Theory and Methods*. Springer-Verlag, New York, 2-nd edition, 1991 (corrected 2-nd printing 1993).
- [4] Peter J. Brockwell and Richard A. Davis. *ITSM for Windows. A User's Guide to Time Series Modelling and Forecasting*. Springer-Verlag, New York, 1994. 2 diskettes included.
- [5] James D. Hamilton. *Time Series Analysis*. Princeton University Press, Princeton, NJ 08540, 1994.
- [6] Lennart Ljung. *SYSTEM IDENTIFICATION: Theory for the User*. Prentice Hall, Inc., Englewood Cliffs, NJ, 1987.

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