

SSA change-point detection for environmental data and monitoring the quality of photovoltaic modules

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- Algorithm of SSA change-point detection
- Possibilities with examples
- ARL in the presence of serial correlation
- Analysis of environmental data
- Application in photovoltaics

Overview of change-point methods

- **CUSUM**
Detects only a change in mean
- **Inflection Point Detection**
Assumes that any variation is a change
- **Autoregressive Modeling**
Assumes a specific generating model
- **Mixtures of Gaussians**
Assumes a specific generating model
- **Discrete Cosine Transform**
Finds only global changes
- **Wavelet Analysis**
Too many parameters
- **Singular Spectrum Analysis (SSA)**
Detects a change in 'structure'

SSA change-point algorithm

The main parameter is N , others are L, k, p, q .

Assumptions

- The distance between change-points is at least N .
- The first change-point occurs after N points.
- The parameter N is big enough to estimate a 'structure' of series.

Transformation of a series

Compound vectors X_1, X_2, \dots from a series x_1, x_2, \dots by applying the moving window of length L ,

$$X_1 = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_L \end{pmatrix}, \quad X_2 = \begin{pmatrix} x_2 \\ x_3 \\ \vdots \\ x_{L+1} \end{pmatrix}, \quad X_3 = \begin{pmatrix} x_3 \\ x_4 \\ \vdots \\ x_{L+2} \end{pmatrix}, \dots$$

We say that a series has a structure of order k if $\dim \mathcal{L}(X_1, X_2, \dots, X_n) = k$ for all $L, n > k$.

Examples of structured series

The series $x_j = A \sin(\gamma j + \omega)$ has a structure of order 2 since

$$X_{j+1} = \cos(\gamma j) \begin{pmatrix} \sin(\gamma + \omega) \\ \sin(2\gamma + \omega) \\ \vdots \\ \sin(L\gamma + \omega) \end{pmatrix} + \sin(\gamma j) \begin{pmatrix} \cos(\gamma + \omega) \\ \cos(2\gamma + \omega) \\ \vdots \\ \cos(L\gamma + \omega) \end{pmatrix}.$$

The series $x_j = \sum_{i=1}^m A_i e^{-\beta_i j} \sin(\gamma_i j + \omega_i)$ has a structure of order $2m$.

SSA estimation of a structure

- 1) For a noisy series $(x_{n+1}, \dots, x_{n+N})$, we build vectors $X_{n+1}, \dots, X_{n+N-L+1}$.
- 2) Make the SVD decomposition $\mathbb{X} = \sum_{i=1}^L \sqrt{\lambda_i} U_i V_i^T$, $\lambda_1 \geq \lambda_2 \geq \dots$.
- 3) Select k principal components with largest λ_i . Components with small λ_i correspond to a noise.
- 4) Define the subspace as $\mathcal{L}(U_1, U_2, \dots, U_k)$, which describes a structure of series.

Informal description of a change-point

We say that there is a change-point in a series

$$\dots, \underbrace{x_{n+1}, \dots, x_{n+N}}_{\text{base series}}, \dots, \underbrace{x_{n+p+1}, \dots, x_{n+q+L-1}}_{\text{test series}}, \dots$$

if the 'test' series $x_{n+p}, \dots, x_{n+q+L-1}$ does not share the structure of the 'base' series x_{n+1}, \dots, x_{n+N} .

In this case,
the change-point belongs to $[n + N, n + q + L - 1]$.

Formal description of a change-point

Consider the statistic $D_{n,k,p,q}$ defined as a distance between vectors $X_{n+p+1}, \dots, X_{n+q}$ and the subspace $\mathcal{L}(U_1, U_2, \dots, U_k)$,

$$D_{n,k,p,q} = \frac{1}{L(q-p)} \sum_{i=n+p+1}^{n+q} [X_i^T X_i - X_i^T U U^T X_i]$$

where $U = (U_1, \dots, U_k)$ is a 'structure' of the series x_{n+1}, \dots, x_{n+N} .

Rule.

There is a change-point in a series if $D_{n,k,p,q} > h$.

Asymptotic behaviour

Theorem. [Moskvina, Zhigljavsky (2003)]

Under certain assumptions, we have

$$\frac{D_{n,k,p,q} - a}{s} \approx N(0, 1)$$

where

$$a = \mathbf{E}D_{n,k,p,q} = \sigma^2 LQ, \quad Q = q - p,$$

and

$$s^2 = \text{Var}D_{n,k,p,q} = \sigma^2 \frac{4}{3} Q(3LQ - Q^2 + 1).$$

Final step of algorithm

Define the normalized statistic

$$S_n = D_{n,k,p,q} / D_{n,k,0,N-L}.$$

Consider the process $W_1 = 0, W_2, W_3, \dots$ defined as

$$W_{n+1} = \max \left\{ W_n + S_{n+1} - S_n - 1/(3LQ), 0 \right\}.$$

Rule.

The point $\tau = n + q + L - 1$ is a change-point if $W_n > h$,

$$h = \frac{2t_\alpha}{LQ} \sqrt{Q(3LQ - Q^2 + 1)/3}$$

and t_α is the $(1 - \alpha)$ -quantile of the standard normal d.

Choice of parameters

- 1) N should be large enough to sufficiently well estimate a 'structure' of series.
- 2) Set $L = N/2$, $p = N$, $q = N + 1$.
- 3) Estimate k from all available data.

Thus, we have the situation where

$$\dots, \underbrace{x_{n+1}, \dots, x_{n+N}}_{\text{base series}}, \underbrace{x_{n+N+1}, \dots, x_{n+N+L}}_{\text{test series}}, \dots$$

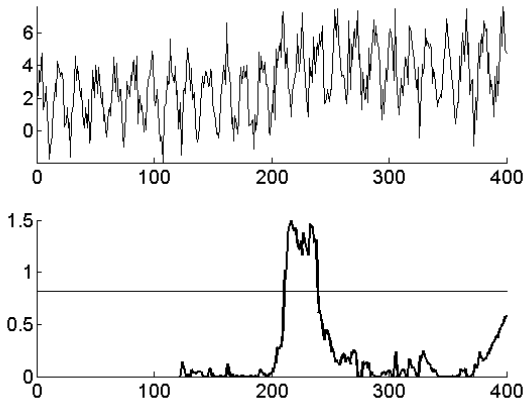
or, alternatively,

$$\underbrace{X_{n+1}, \dots, X_{n+N-L+1}}_{\text{base vectors}}, \underbrace{X_{n+N+1}}_{\text{test vector}}$$

Change in mean

$$x_n = 2 + 2 \sin(0.4n) + \varepsilon_n \text{ for } n \leq 200 \text{ and}$$

$$x_n = 4 + 2 \sin(0.4n) + \varepsilon_n \text{ for } n > 200, \varepsilon_n \sim N(0, 1)$$

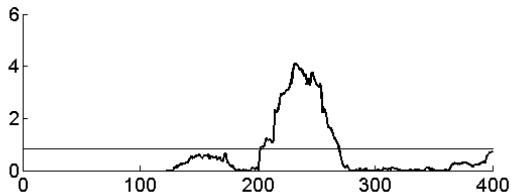
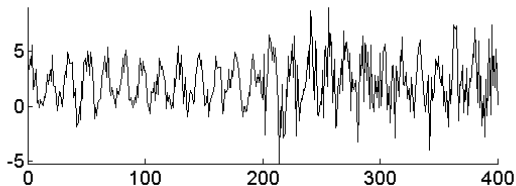


$$N = 80, k = 3$$

Change in variance

$$x_n = 2 + 2 \sin(0.4n) + \varepsilon_n \text{ for } n \leq 200 \text{ and}$$

$$x_n = 2 + 2 \sin(0.4n) + 2\varepsilon_n \text{ for } n > 200, \varepsilon_n \sim N(0, 1)$$

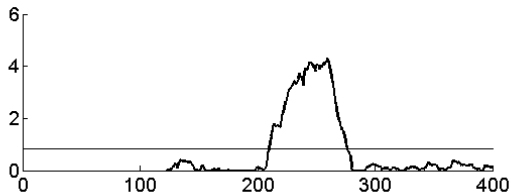
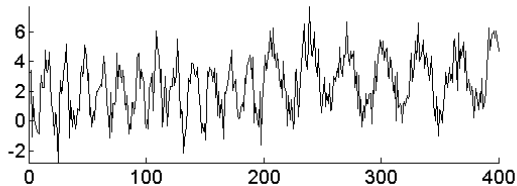


$$N = 80, k = 3$$

Change in frequency

$$x_n = 2 + 2 \sin(0.4(n - 200)) + \varepsilon_n \text{ for } n \leq 200 \text{ and}$$

$$x_n = 2 + 2 \sin(0.2n) + \varepsilon_n \text{ for } n > 200, \varepsilon_n \sim N(0, 1)$$

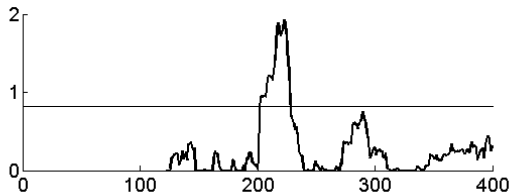
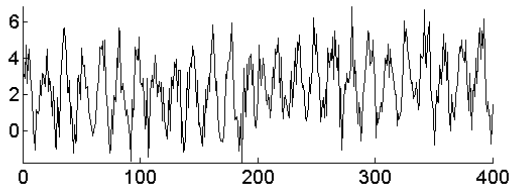


$$N = 80, k = 3$$

Change in phase

$$x_n = 2 + 2 \sin(0.4n) + \varepsilon_n \text{ for } n \leq 200 \text{ and}$$

$$x_n = 2 + 2 \sin(0.4n + 1) + \varepsilon_n \text{ for } n > 200, \varepsilon_n \sim N(0, 1)$$



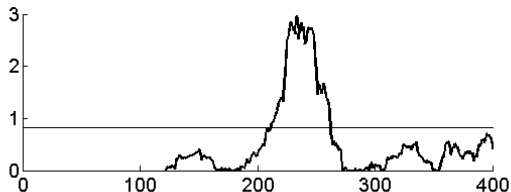
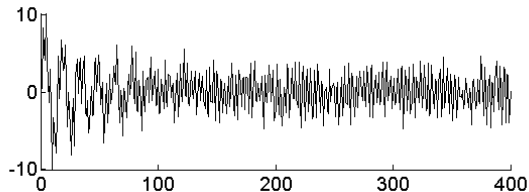
$$N = 80, k = 3$$

Change in AR model with iid noise

$$z_n = -0.96z_{n-4} + z_{n-3} - 0.5z_{n-2} + 0.97z_{n-1}, \quad n = 5, \dots, 200,$$

$$z_n = -0.96z_{n-4} + z_{n-3} - 0.7z_{n-2} + 0.97z_{n-1}, \quad n = 201, \dots, 400,$$

$$x_n = z_n + \varepsilon_n, \quad z_1 = 0, \quad z_2 = 8, \quad z_3 = 6, \quad z_4 = 4$$

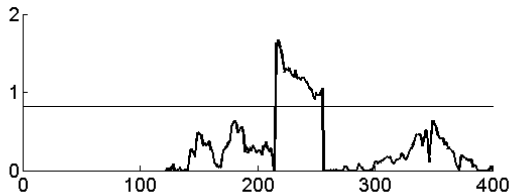
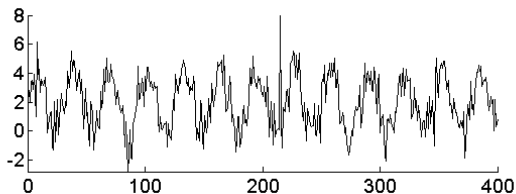


$$N = 80$$

$$k = 4$$

Detection of outliers

$$x_n = 2 + 2 \sin(0.2n) + \varepsilon_n \text{ for } n \neq 215, x_{215} = 8, \\ \varepsilon_n \sim N(0, 1)$$



$$N = 80, k = 3$$

Average run length for AR(1) and MA(1)

SSA change-point detection with $k = 1$, $N = 80$

$$x_n = \mu + \phi(x_{n-1} - \mu) + \varepsilon_n, \quad \varepsilon_n \sim N(0, 1),$$

$$\mu = 0.5$$

ϕ	-0.5	-0.4	-0.2	0	0.2	0.4	0.5
ARL	308	391	498	524	460	224	192

$$x_n = \mu + \varepsilon_n - \theta\varepsilon_{n-1}, \quad \varepsilon_n \sim N(0, 1),$$

$$\mu = 0.5$$

θ	-0.5	-0.4	-0.2	0	0.2	0.4	0.5
ARL	272	302	441	524	512	503	430

Bagshaw M., Johnson R. A. (1974,1975)

The Effect of Serial Correlation on the Performance of CUSUM Tests

Average run length for AR(1) and MA(1)

SSA change-point detection with $k = 3$, $N = 80$

$$x_n = \mu + \phi(x_{n-1} - \mu) + \varepsilon_n, \quad \varepsilon_n \sim N(0, 1),$$

$$\mu = 2 + 2 \sin(0.4n)$$

ϕ	-0.5	-0.4	-0.2	0	0.2	0.4	0.5
ARL	209	333	412	450	355	162	115

$$x_n = \mu + \varepsilon_n - \theta\varepsilon_{n-1}, \quad \varepsilon_n \sim N(0, 1),$$

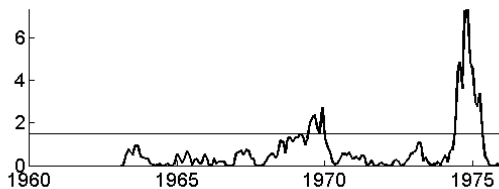
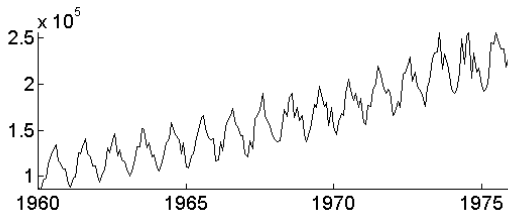
$$\mu = 2 + 2 \sin(0.4n)$$

θ	-0.5	-0.4	-0.2	0	0.2	0.4	0.5
ARL	190	243	337	450	423	415	356

Example: gasoline demand

Abraham, Redolter (1983) Statistical Methods for Forecasting, Wiley
<http://robjhyndman.com/TSDL/sales/>

Monthly gasoline demand in Ontario from 1960 to 1975



$$N = 24$$

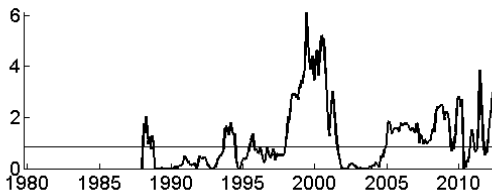
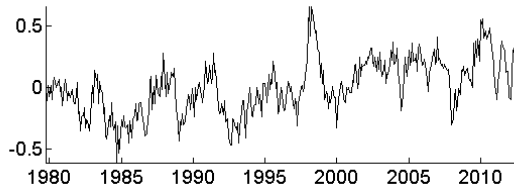
$$k = 5$$

Example: global Earth temperature

National Space Science and Technology Center, USA, NASA

<http://vortex.nsstc.uah.edu/data/msu/t21t/uahncdc.lt>

Monthly Earth temperatures from Dec 1978 to Jun 2012

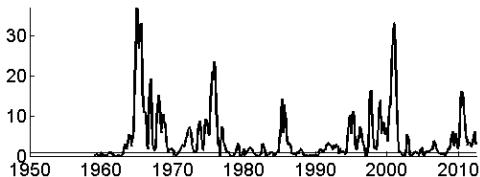
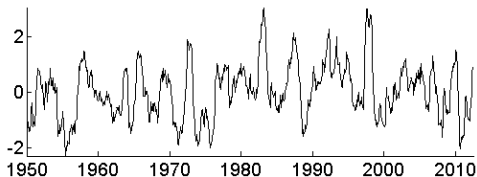


$$N = 72$$

$$k = 9$$

Example: MEI index

Multivariate ENSO Index (MEI) is based on the six main observed variables over the tropical Pacific, from Jan 1950 to Jun 2012, <http://www.esrl.noaa.gov/psd/enso/mei/table.html>



$$N = 72$$

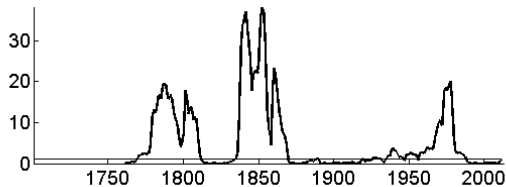
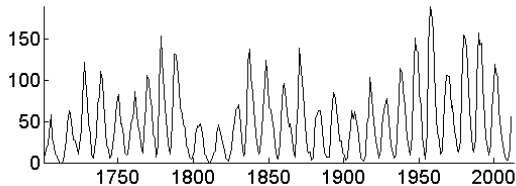
$$k = 8$$

ENSO stands for El Niño-Southern Oscillation

Example: yearly sunspot number

Solar Influences Data Analysis Center, Observatory of Belgium, <http://sidc.oma.be/sunspot-data/>

Yearly sunspot number from 1700 to 2011

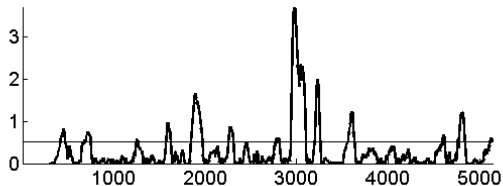
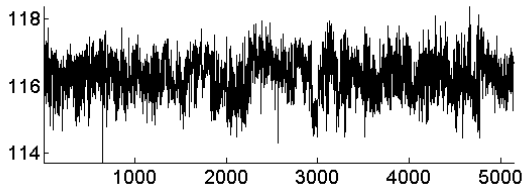


$$N = 40$$

$$k = 4$$

Power measurements of PV modules

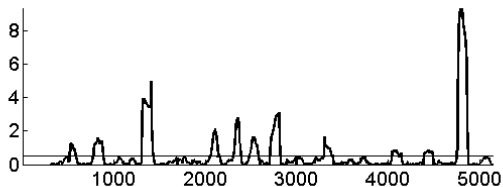
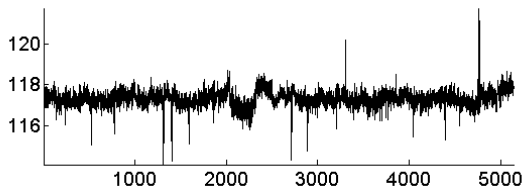
Power of PV modules obtained by using a flasher (sun simulator) from a production line 1



$$N = 200$$
$$k = 1$$

Power measurements of PV modules

Power of PV modules obtained by using a flasher (sun simulator) from a production line 2



$$N = 200$$

$$k = 1$$

Thank you for your attention!