

9th EUMETNET Data Management Workshop

6th-8th November 2013, El Escorial, Madrid



Detection of inhomogeneities in daily data: a test based in the Kolmogorov-Smirnov test

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- Definitions
- Control Analysis
- Inhomogeneity detection

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- ♦ Temperature
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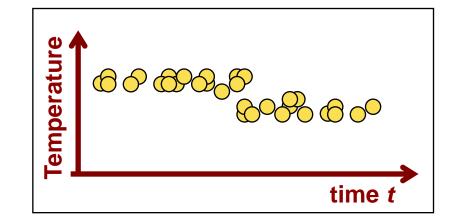
Introduction

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Introduction

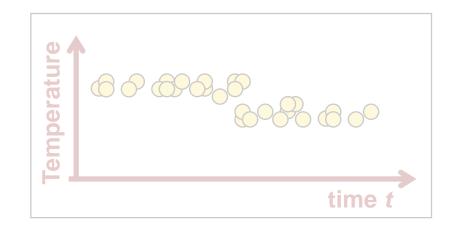
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Problem: Abrupt change in observations

- Changes in the weather shelter (best / worst ventilation, white painting, ...)
- Changes in meteorological sensors
- Changes in the location of weather stations
- Changes in the environment (vegetation, buildings, ...)





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Current solutions

- SNHT: Standard Normal Homogeneity Test (by Alexandersson, 1986)
- Others methods at **monthly scale**: Buishand range test, Pettitt test, von Neumann ratio tests,...

- Some methods at **daily scale** (mean, quantiles or moments): Using parallel measurements, reference series,... combining with corrections as HOM, HOMAD, SPLIDHOM, QM, PM,... **but** from a **breakpoint** detection at **monthly** scale.





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Current limitations

- Detection of inhomogeneities in daily time-series is partially saved.
- Alternatives should be explored



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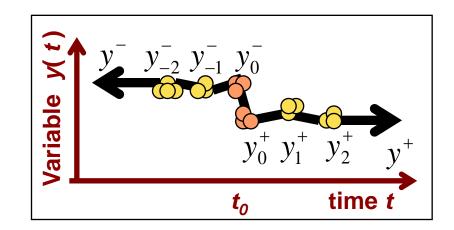
Methodology

Definitions

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- $y_i \equiv$ set of daily data
- $t_0 \equiv$ inhomogeneity candidate
- $t^{-} \equiv$ time values to the left
- $t^+ \equiv$ time values to the right
- PV ≡ p-value of KS test







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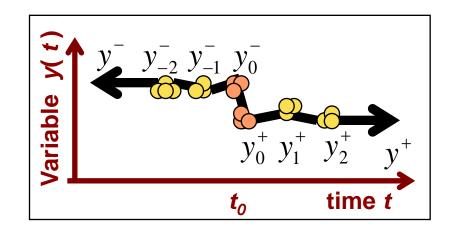
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Measure of the dissimilitude between two sets (e.g. 365 days)

$$LPV_j^i \equiv \log_{10}(PV(y_i, y_j))$$

$$PV_j^i \in (0, 1) \longrightarrow LPV_j^i \in (-\infty, 0)$$

High ↔ low dissimilitude





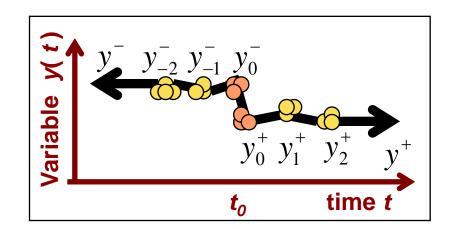
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Measure of possible Inhomogeneity jumps: Dissimilitude between 2 contiguous sets $LPV_{i+1}^{i} \equiv \log_{10}(PV(y_{i}, y_{i+1}))$

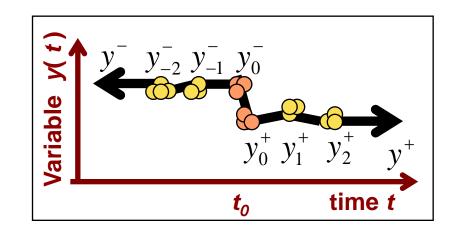






Control Analysis

Introducing an artifitial inhomogeneity to each set of data $y_i \rightarrow \tilde{y}_i$









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$$LPV_{\{i+1\}}^{i} \equiv \log_{10}(PV(y_{i}, \tilde{y}_{i+1}))$$







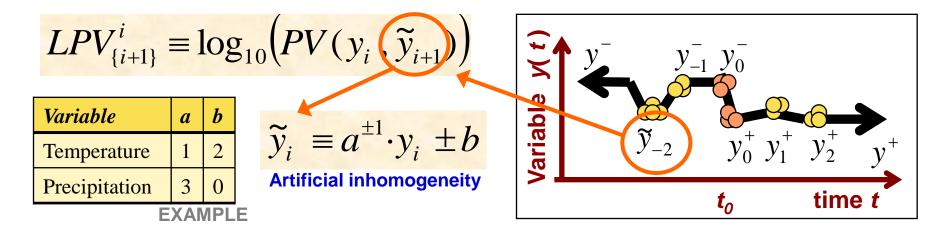
Control Analysis

Methodology

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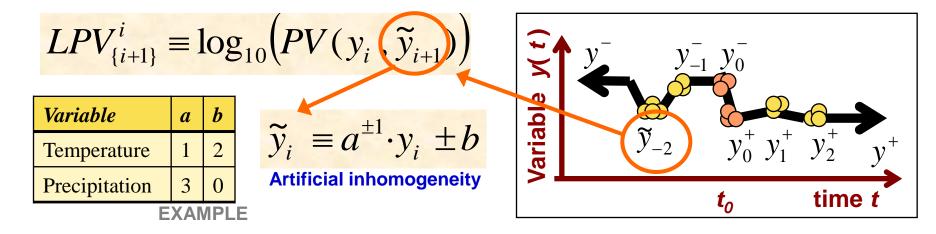


Control Analysis

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Methodology

Introducing an artifitial inhomogeneity to each set of data $y_i \rightarrow \tilde{y}_i$



Reference LPV is defined from the average value: inhomogeneity of control

$$LPV_{Inh} \equiv \frac{1}{N-1} \sum_{i=1}^{N-1} LPV_{\{i+1\}}^{i}$$



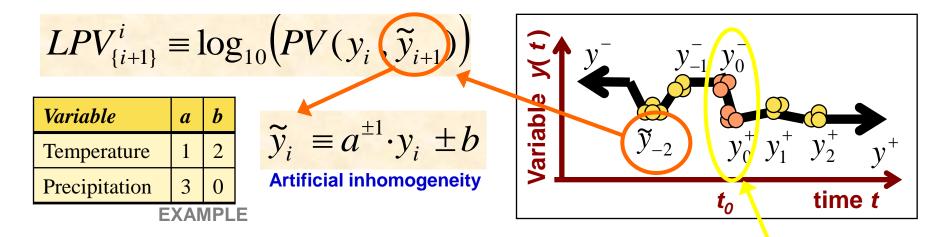




Control Analysis

Methodology

Introducing an artifitial inhomogeneity to each set of data $y_i \rightarrow \tilde{y}_i$



Reference LPV is defined from the **average value**: inhomogeneity of control

$$LPV_{Inh} \equiv \frac{1}{N-1} \sum_{i=1}^{N-1} LPV_{\{i+1\}}^{i} \implies LPV_{i+1}^{i} \leq LPV_{Inh} \rightarrow \begin{cases} y_{0}^{-} = y_{i} \\ y_{0}^{+} = y_{i+1} \end{cases}$$



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 y^+

Inhomogeneity detection

Similarity between the candidates y_0 and the other populations that are on the left (-) and right (+)

$$LPV_{-}^{-} \equiv \log_{10} \left(PV(y^{-}, y_{0}^{-}) \right) \text{ Similar}$$

$$LPV_{-}^{+} \equiv \log_{10} \left(PV(y^{+}, y_{0}^{-}) \right) \text{ different}$$

$$LPV_{+}^{-} \equiv \log_{10} \left(PV(y^{-}, y_{0}^{+}) \right) \text{ different}$$

$$LPV_{+}^{+} \equiv \log_{10} \left(PV(y^{+}, y_{0}^{+}) \right) \text{ Similar}$$

$$t_{0} \text{ time}$$



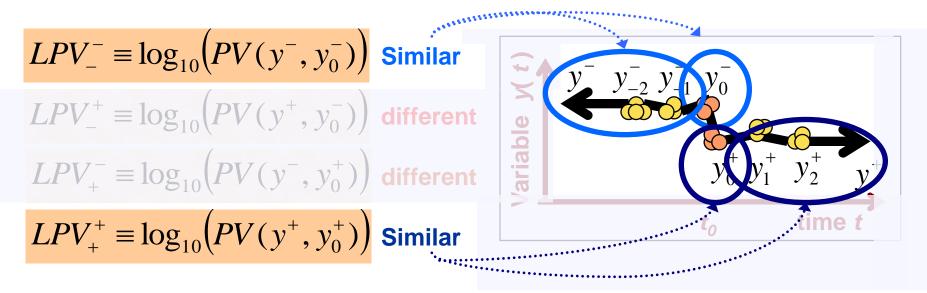
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Inhomogeneity detection

Similarity between the candidates y_0 and the other populations that are on the left (-) and right (+)





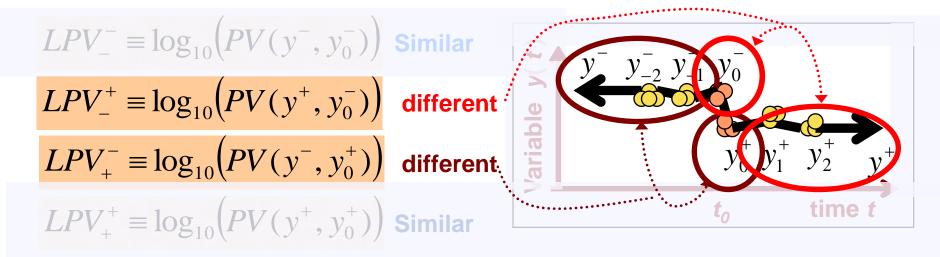
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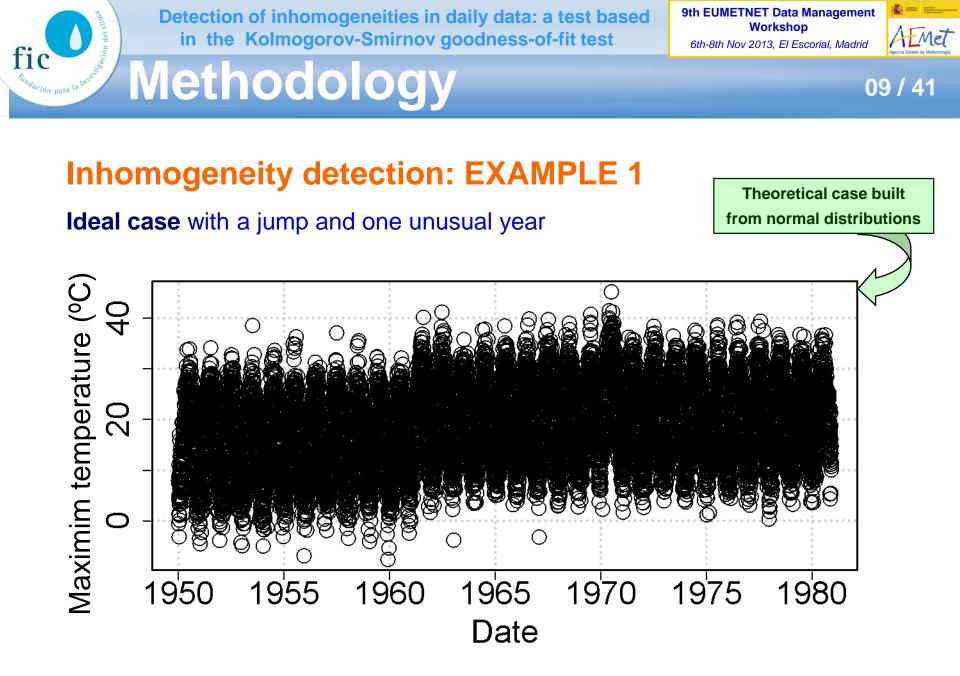


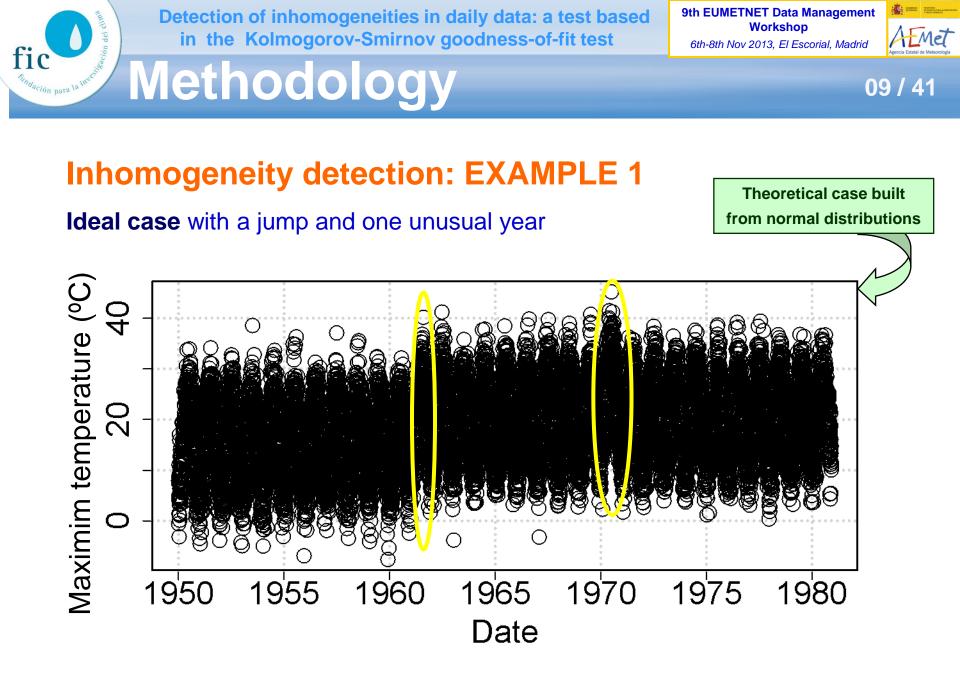


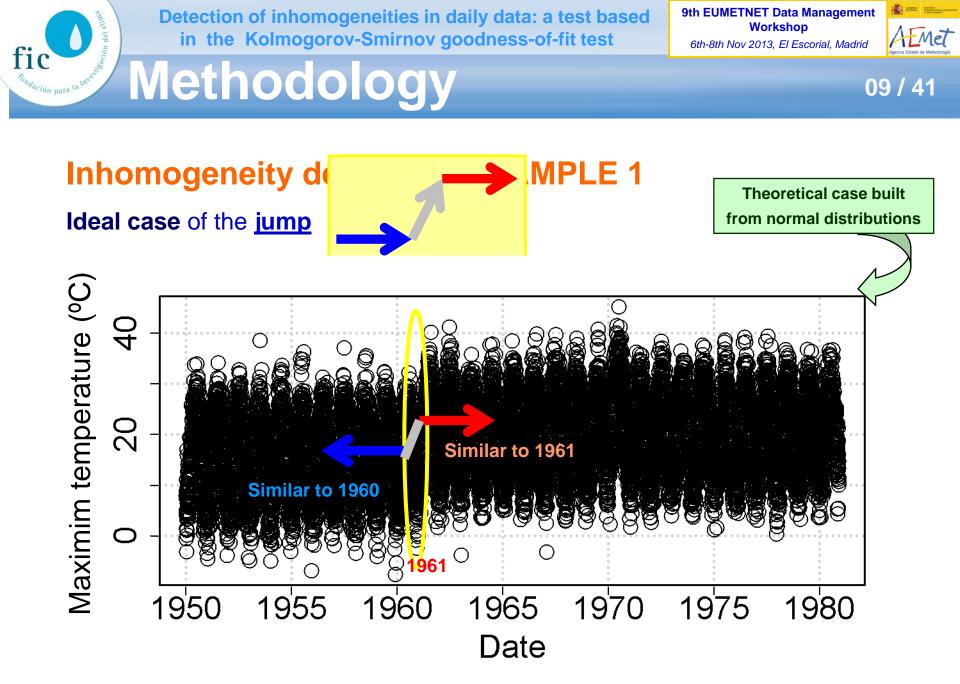
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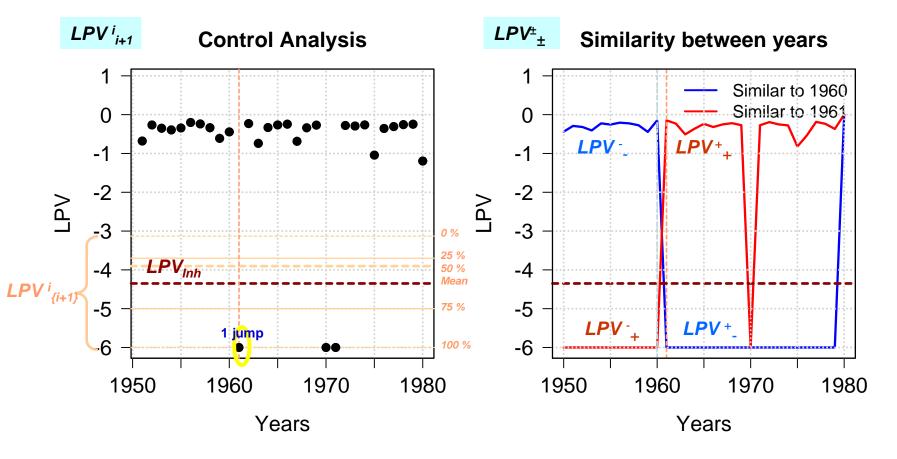
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Inhomogeneity detection: EXAMPLE 1

LPV diagrams for the ideal case of jump





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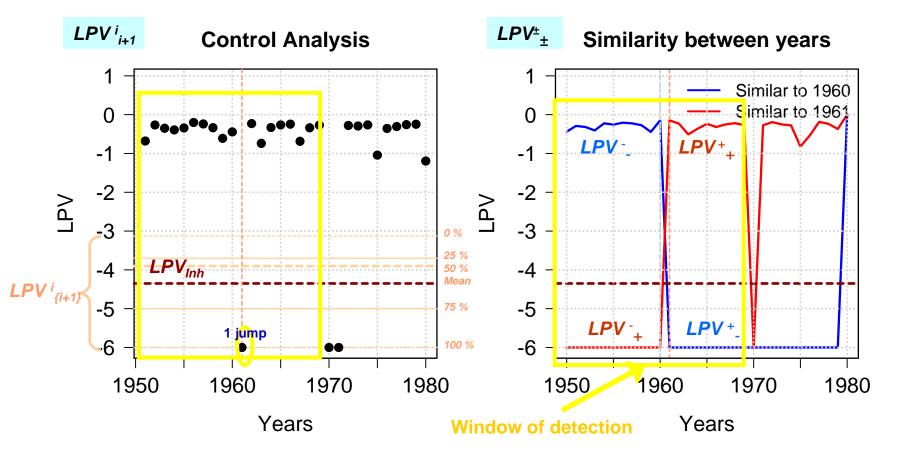
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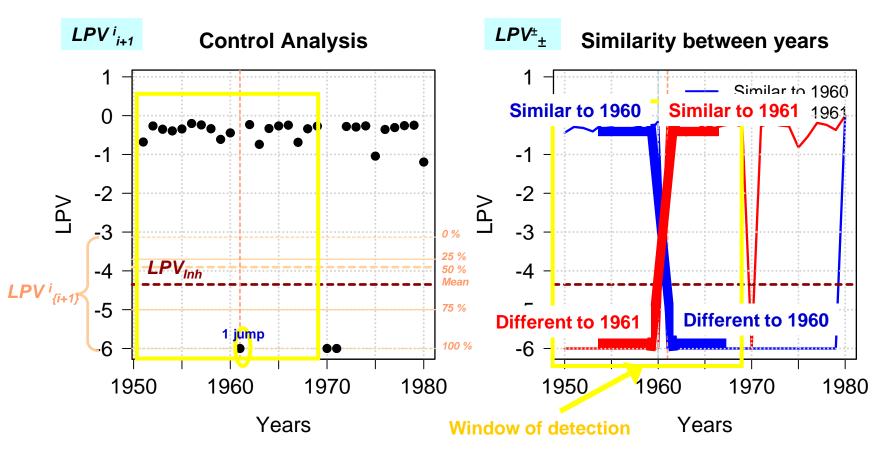
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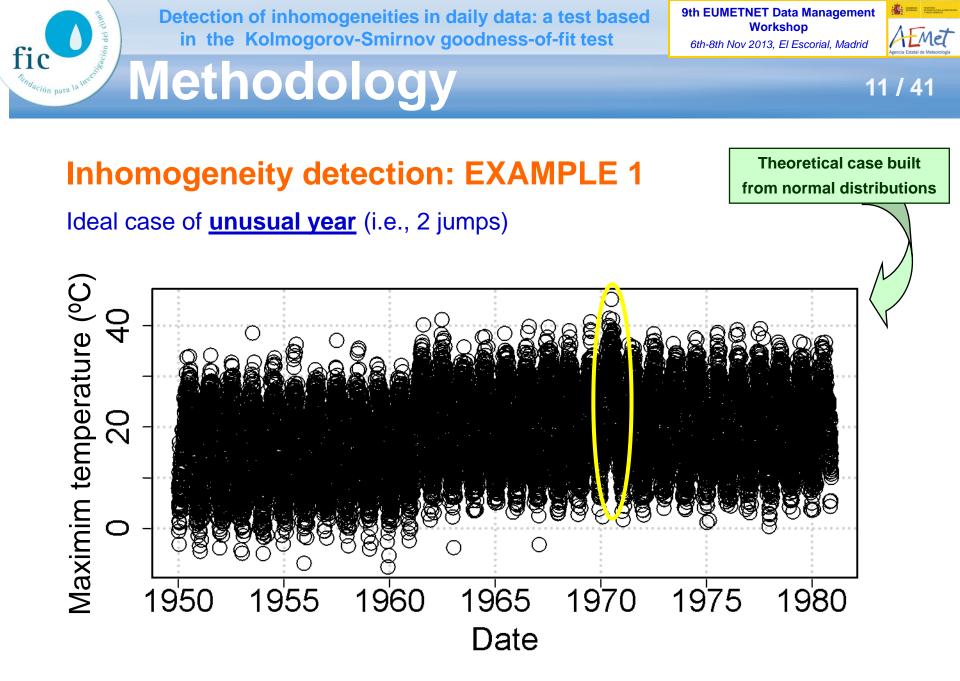


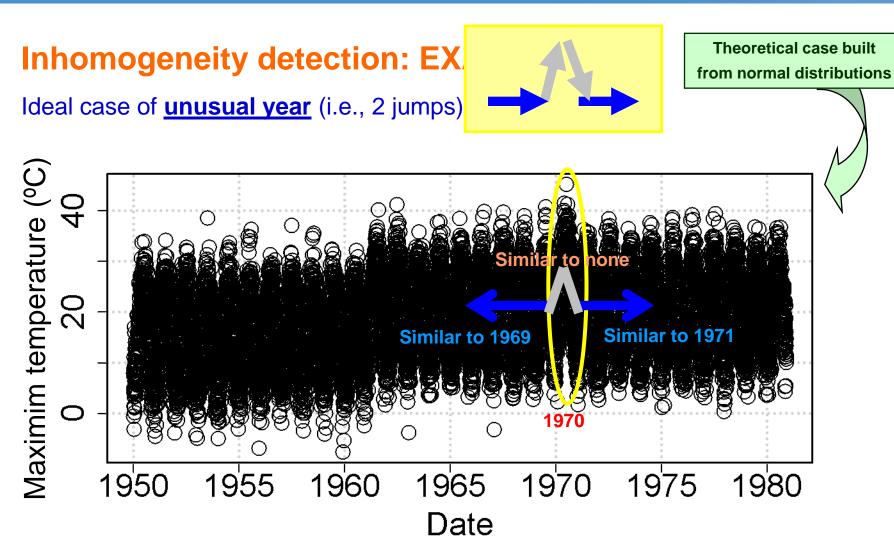


Inhomogeneity detection: EXAMPLE 1

LPV diagrams for the ideal case of jump







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Detection of inhomogeneities in daily data: a test based in the Kolmogorov-Smirnov goodness-of-fit test 9th EUMETNET Data Management Workshop 6th-8th Nov 2013, El Escorial, Madrid



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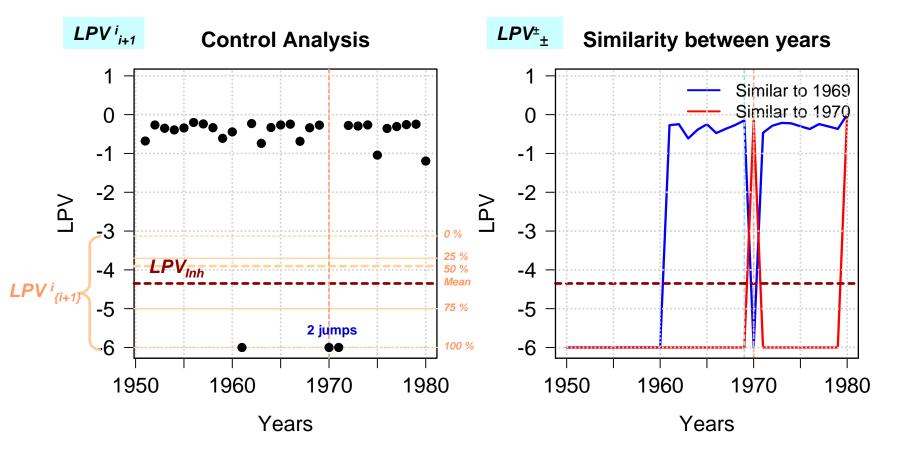






Inhomogeneity detection: EXAMPLE 1

LPV diagrams for the ideal case of <u>unusual year</u> (2 jumps)



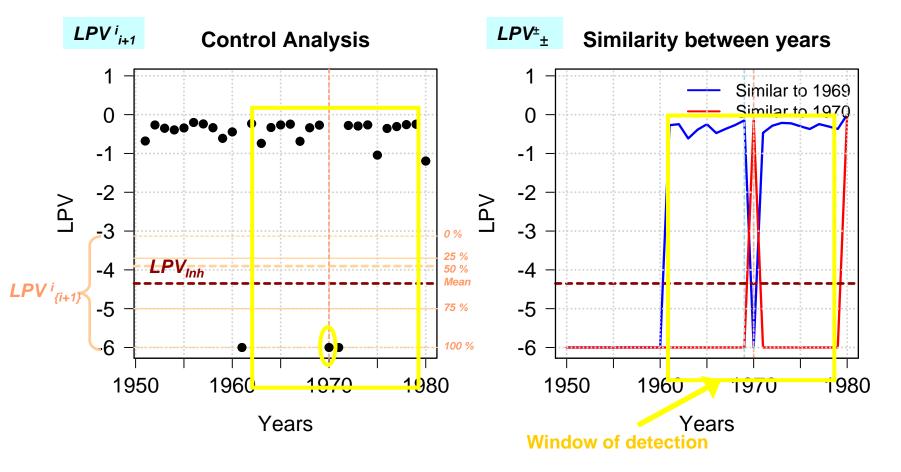






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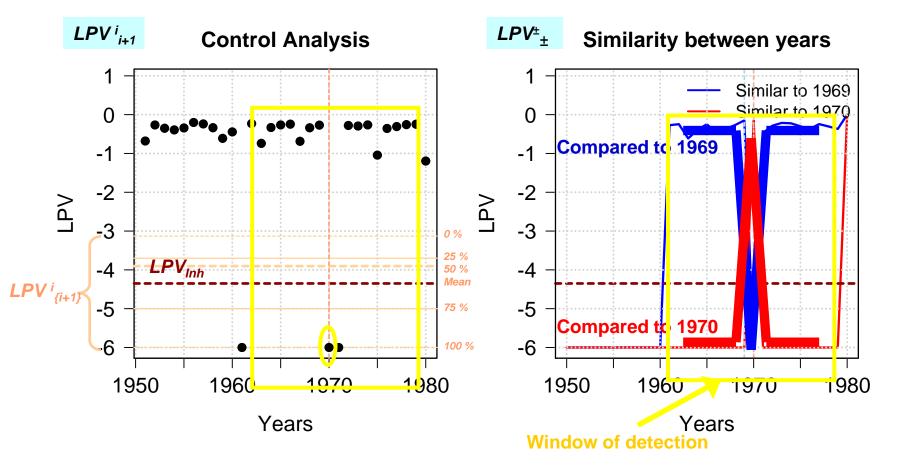


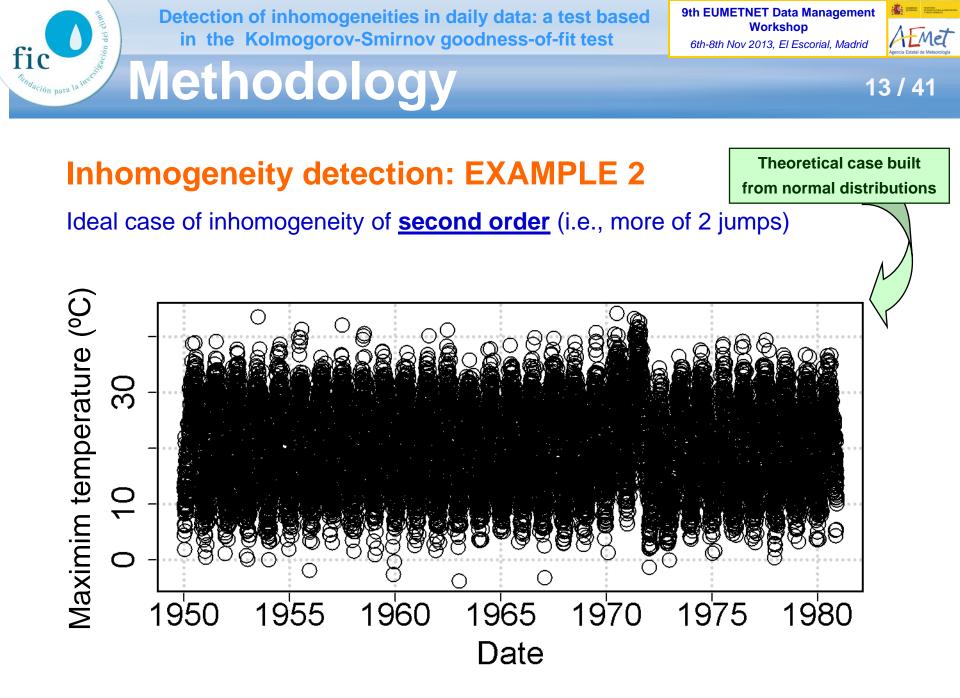


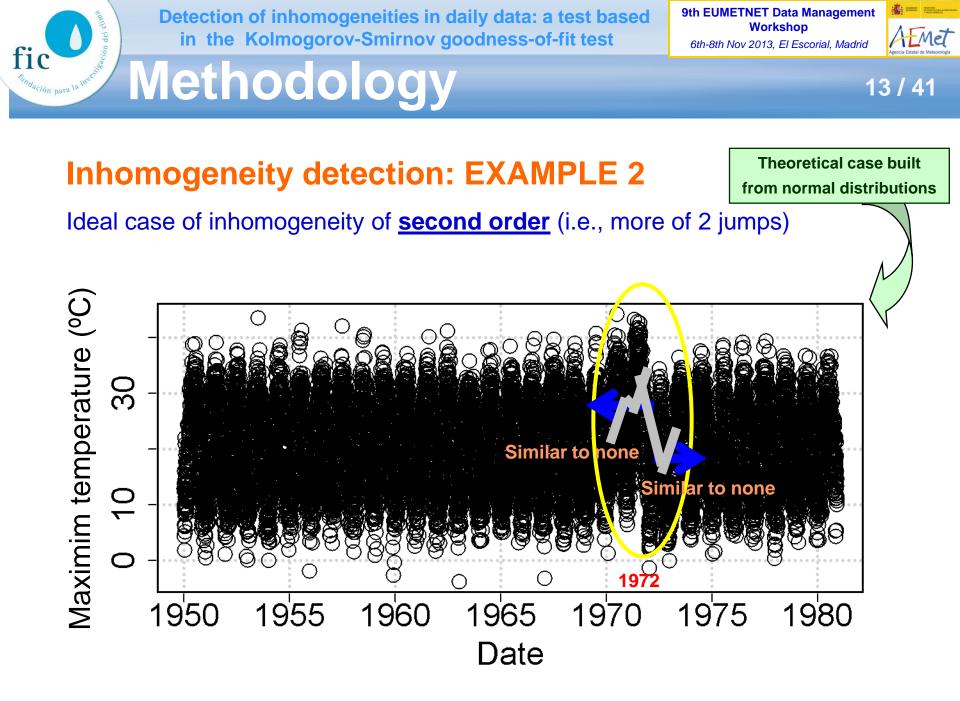


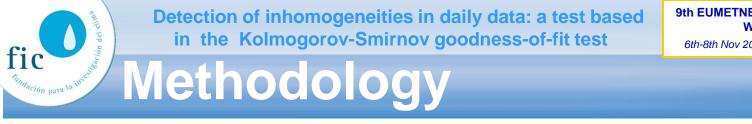
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LPV diagrams for the ideal case of <u>unusual year</u> (2 jumps)







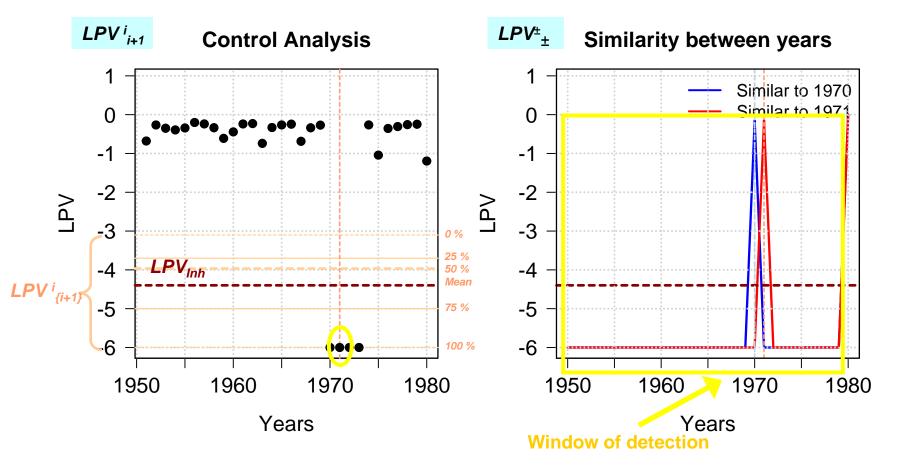






Inhomogeneity detection: EXAMPLE 2

LPV diagrams for the ideal case of inhomogeneity of second order





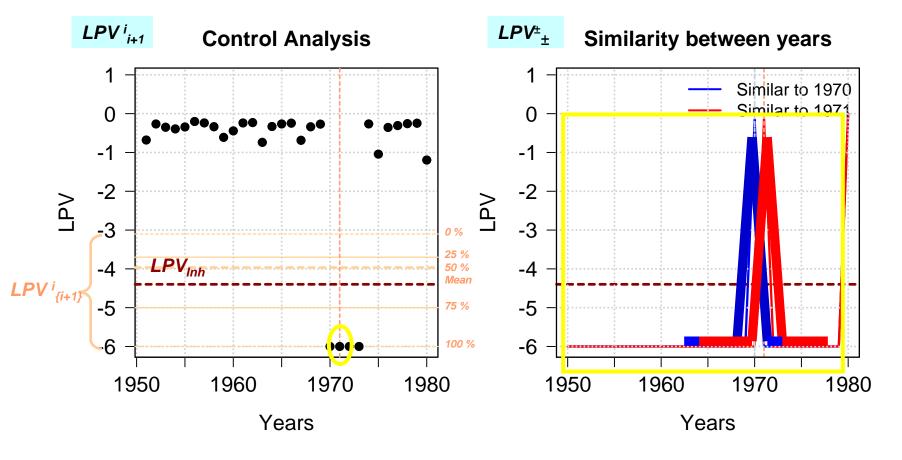
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Inhomogeneity detection: EXAMPLE 2

LPV diagrams for the ideal case of inhomogeneity of second order





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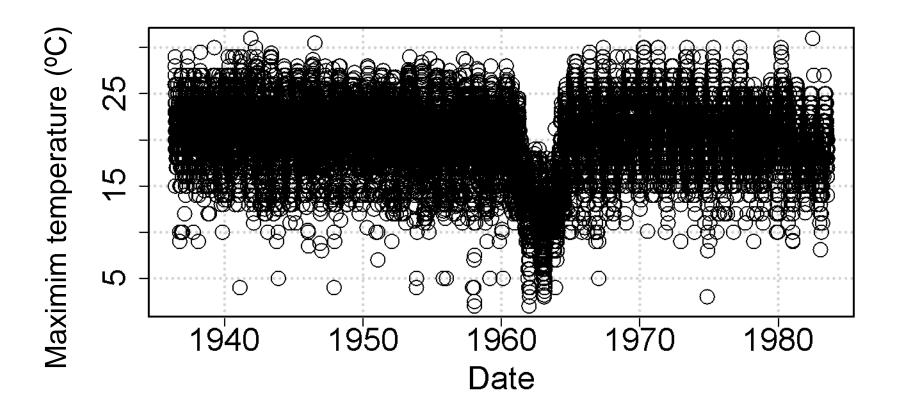
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Temperature: CASE 1

Inhomogeneities in temperature



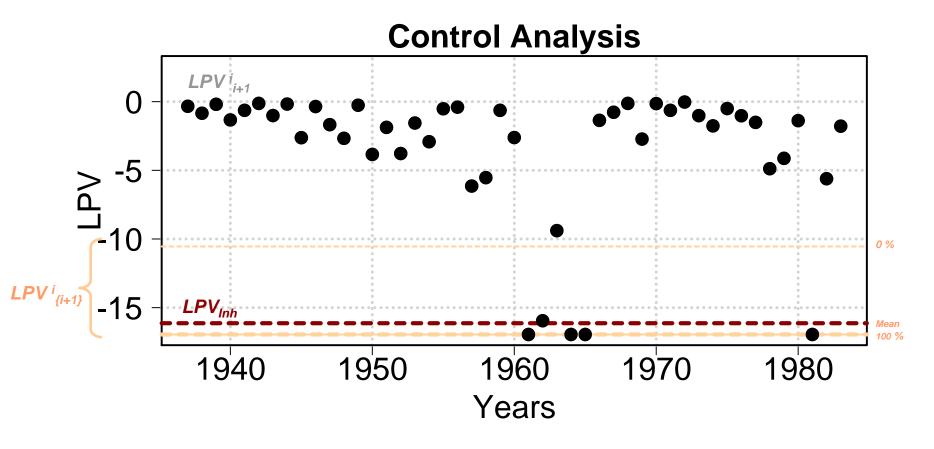


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Temperature: CASE 1



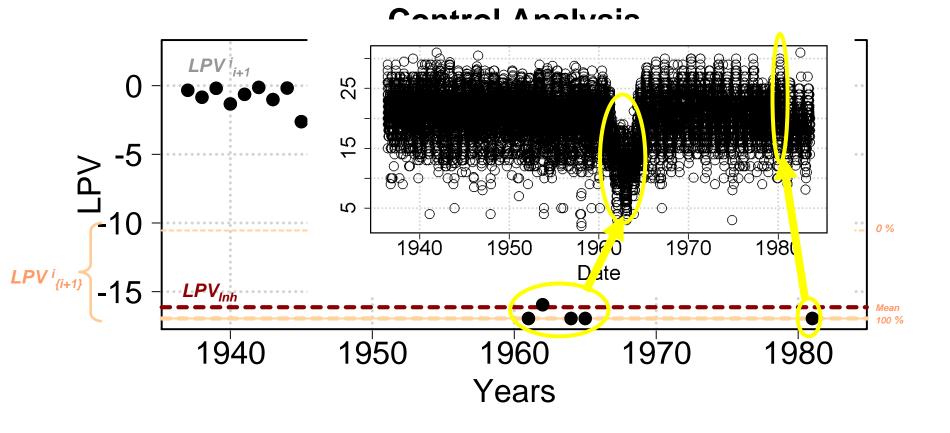


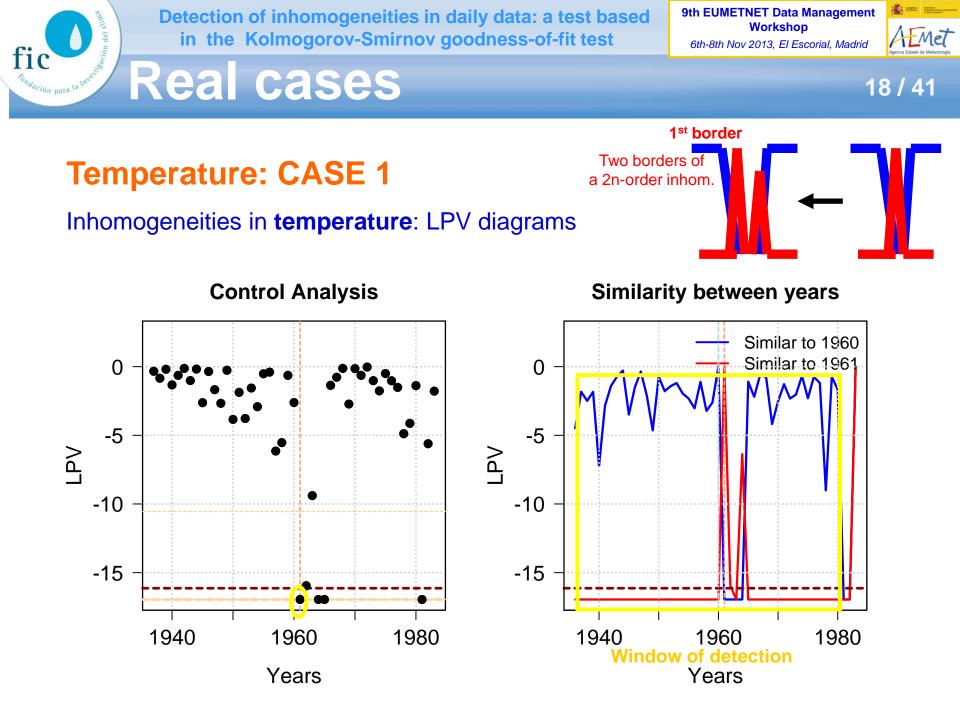
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Temperature: CASE 1





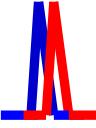


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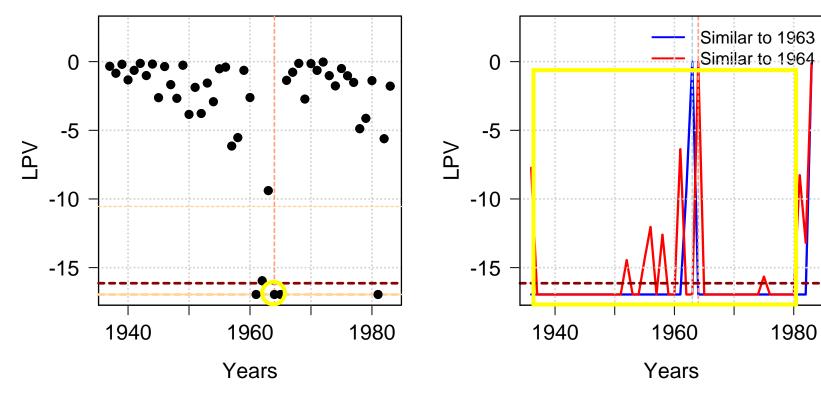


Temperature: CASE 1



Control Analysis

Similarity between years



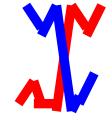


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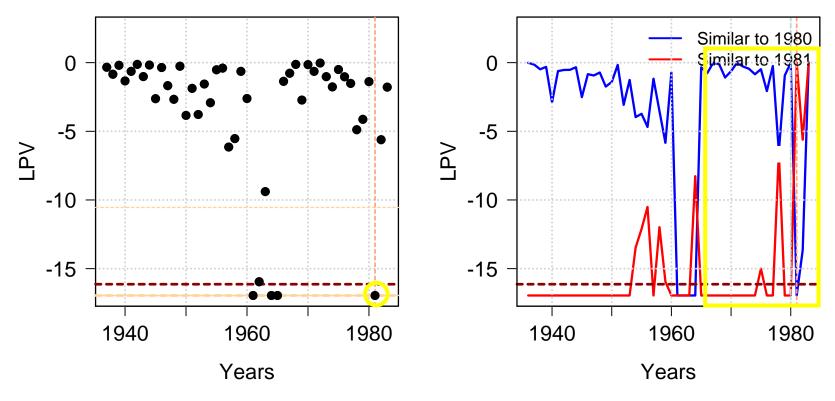
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Temperature: CASE 1



Control Analysis

Similarity between years





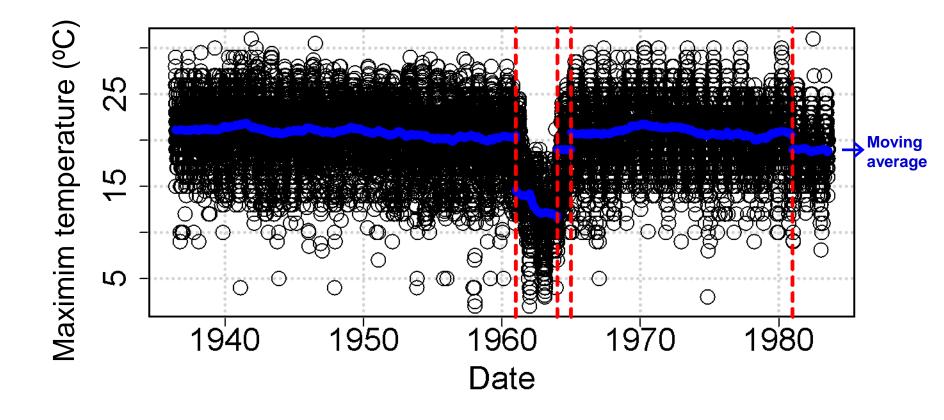
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Temperature: CASE 1

Inhomogeneities in temperature: final analysis





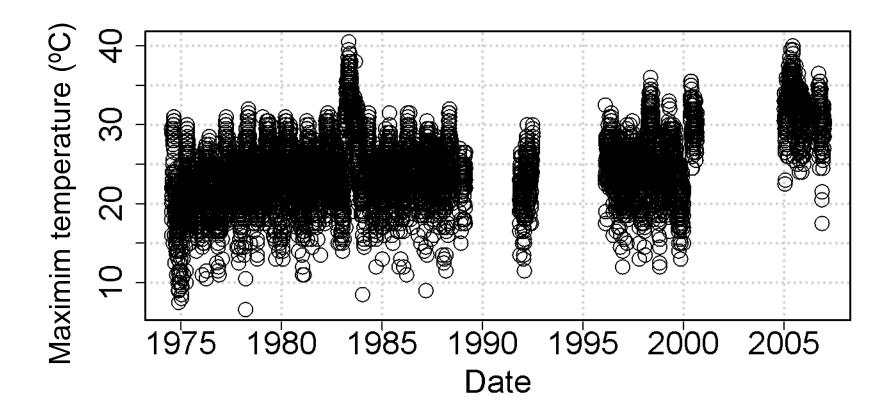
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Temperature: CASE 2

Inhomogeneities in temperature



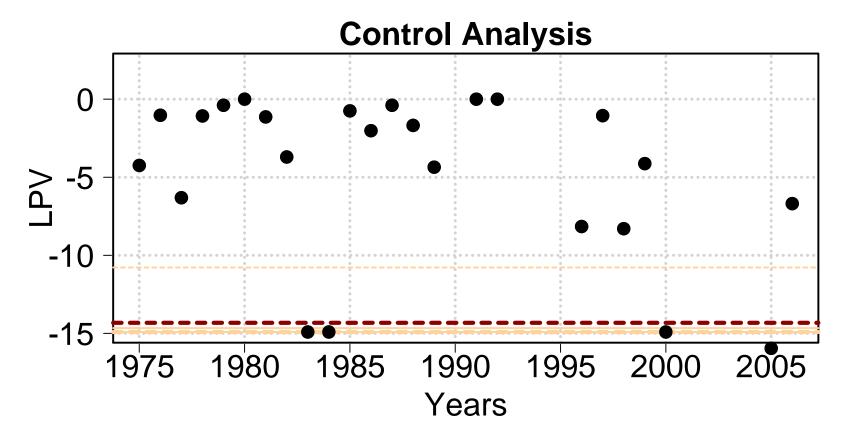


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Temperature: CASE 2



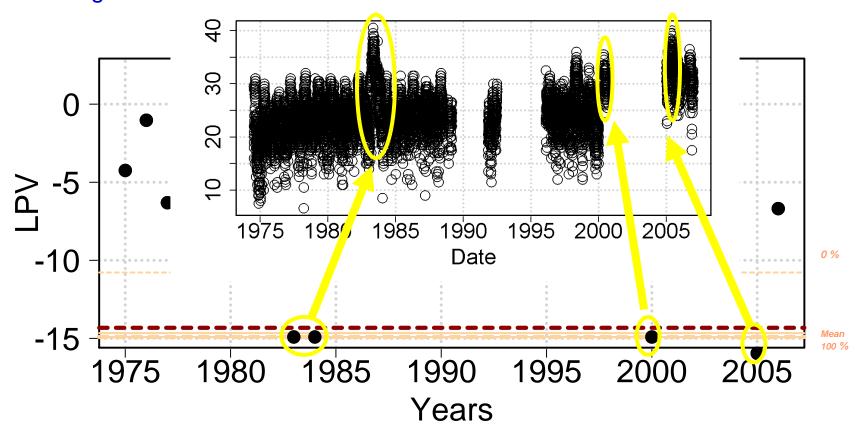


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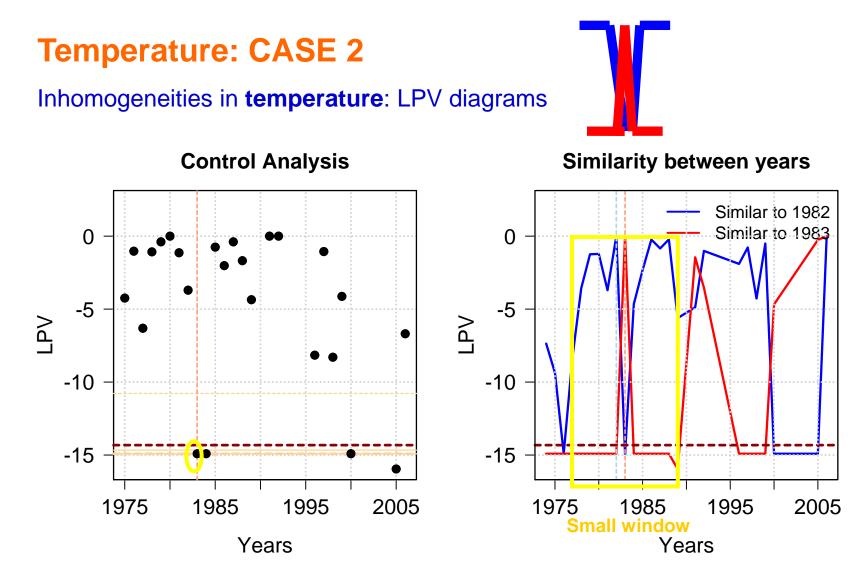
Temperature: CASE 2







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Temperature: CASE 2

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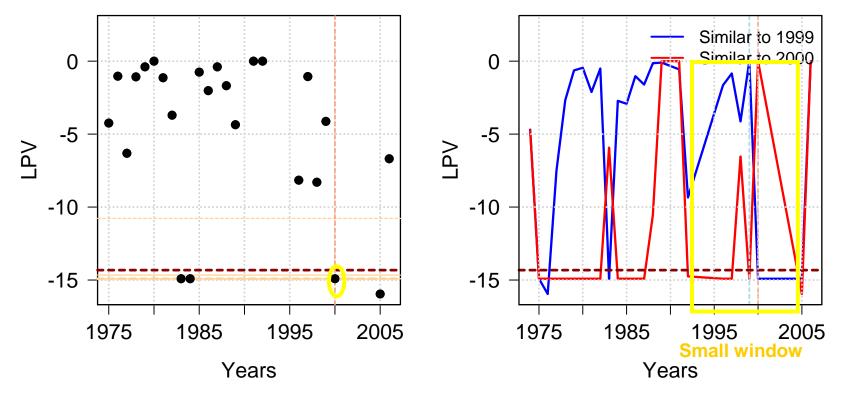
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Control Analysis

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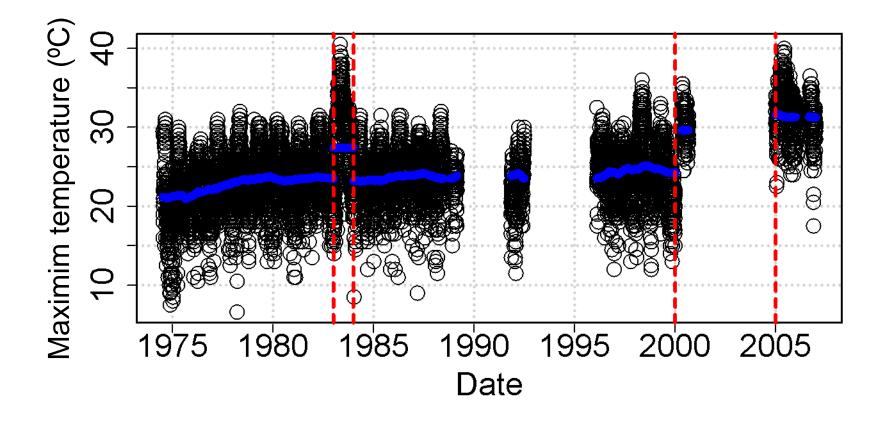
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Temperature: CASE 2

Inhomogeneities in temperature: final analysis





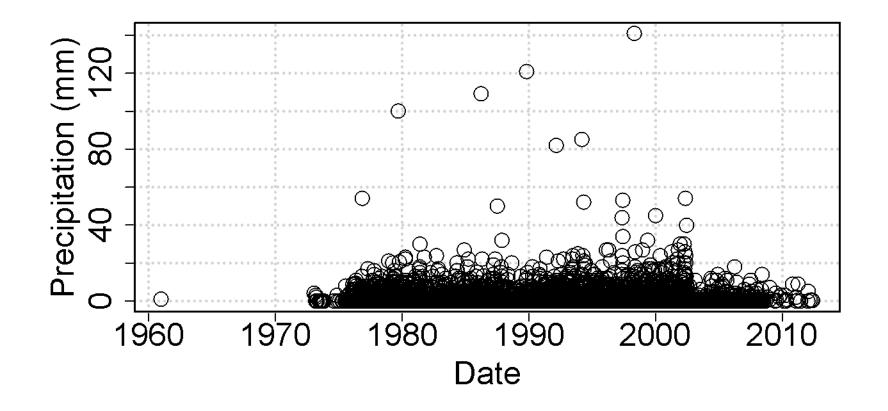
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Precipitation: CASE 1

Inhomogeneities in precipitation



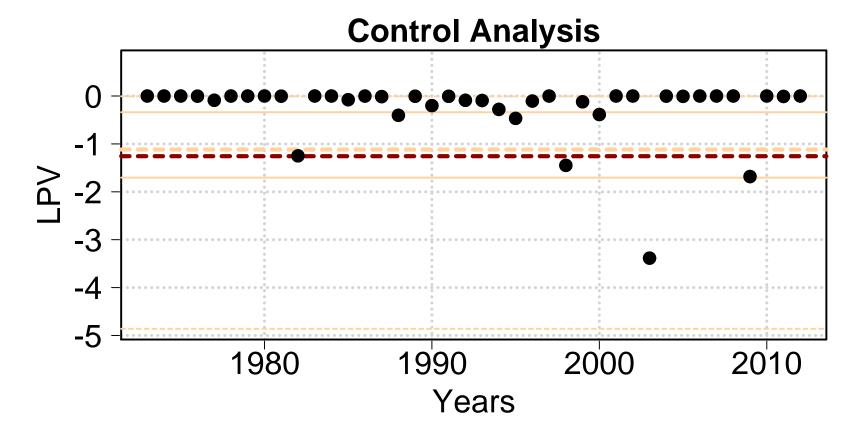


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Precipitation: CASE 1



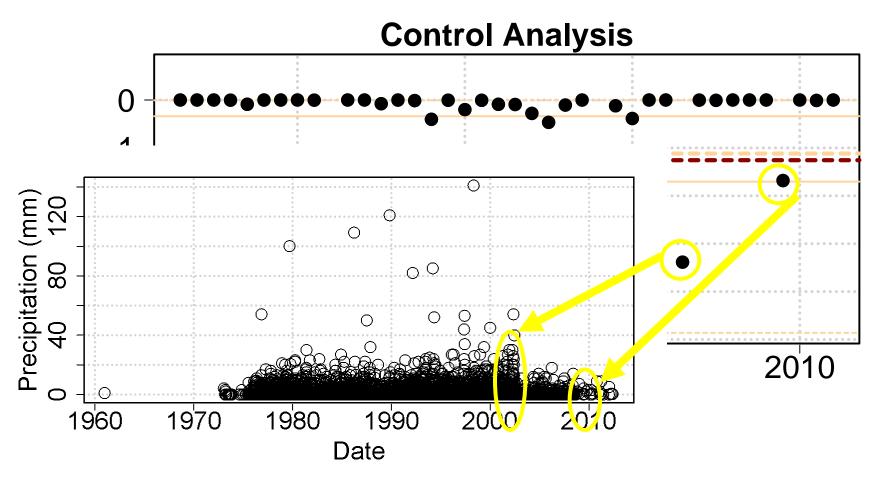


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Precipitation: CASE 1



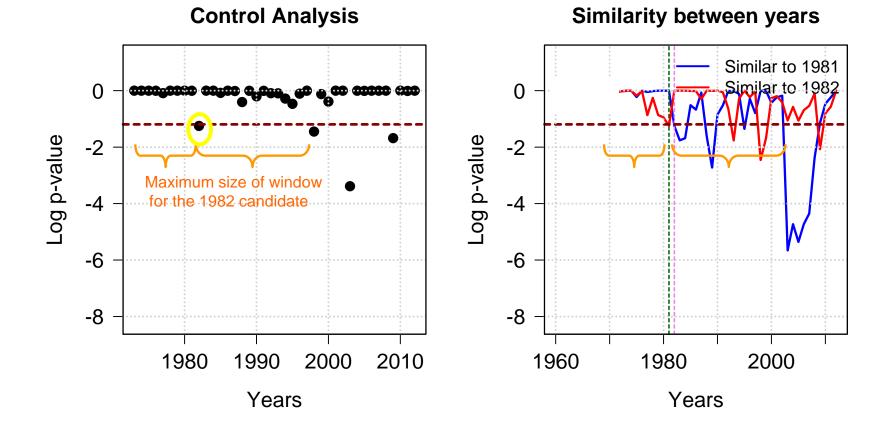


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Precipitation: CASE 1



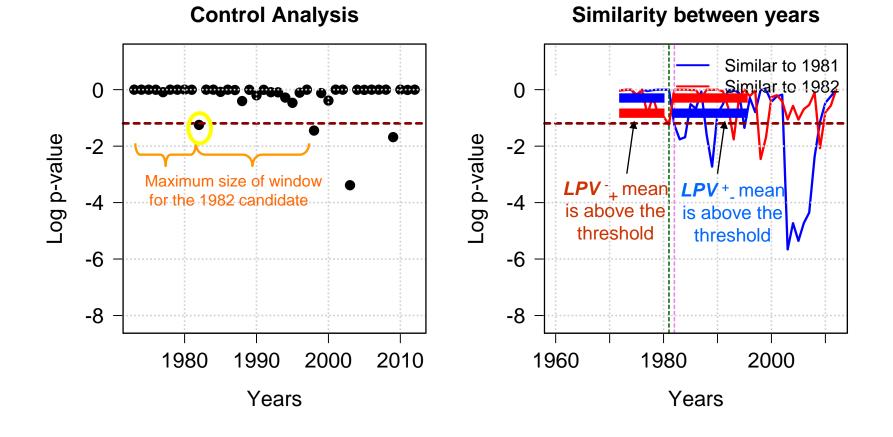


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Precipitation: CASE 1



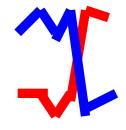


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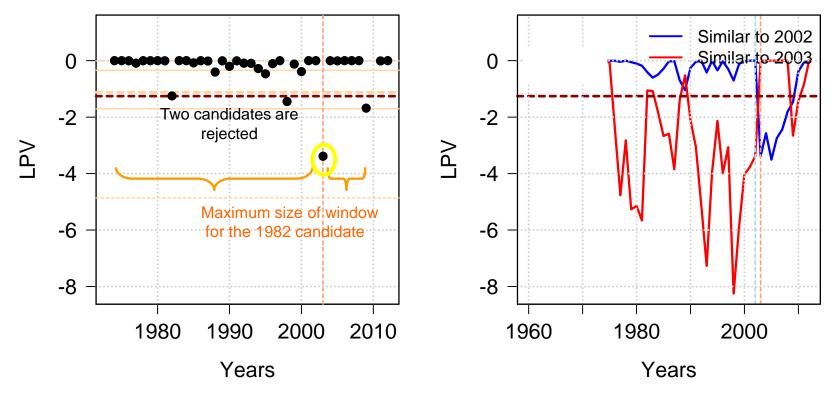
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Precipitation: CASE 1



Control Analysis

Similarity between years





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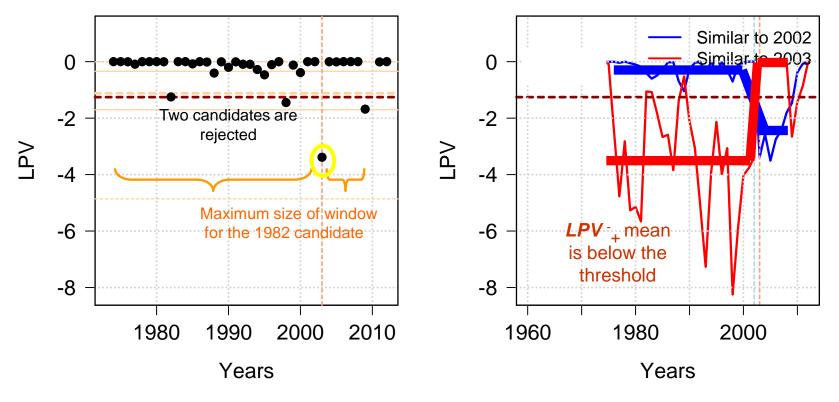
Precipitation: CASE 1

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Control Analysis

Similarity between years





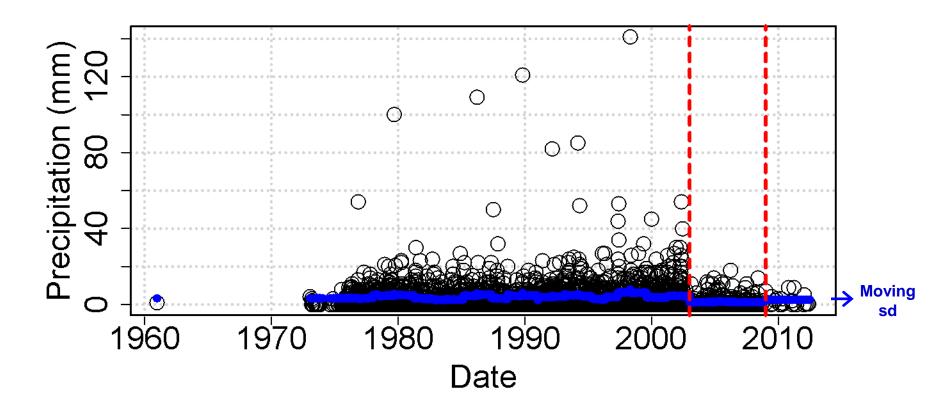
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Precipitation: CASE 1

Inhomogeneities in precipitation: final analysis





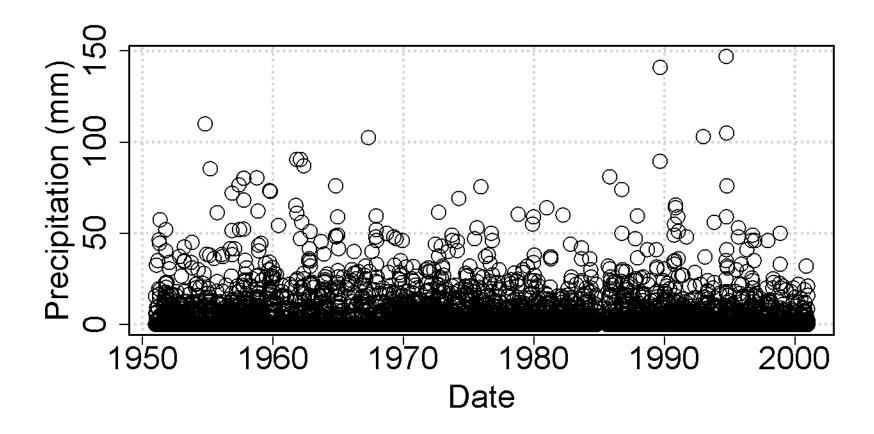
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Precipitation: CASE 2

Inhomogeneities in precipitation





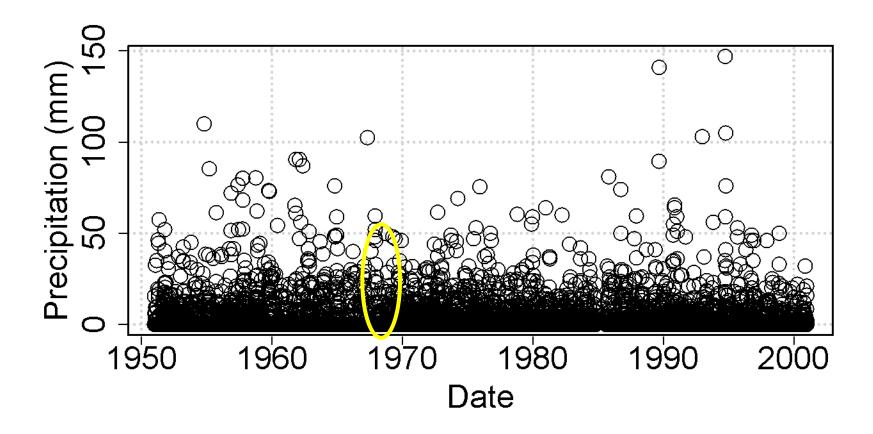
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Precipitation: CASE 2

Inhomogeneities in precipitation



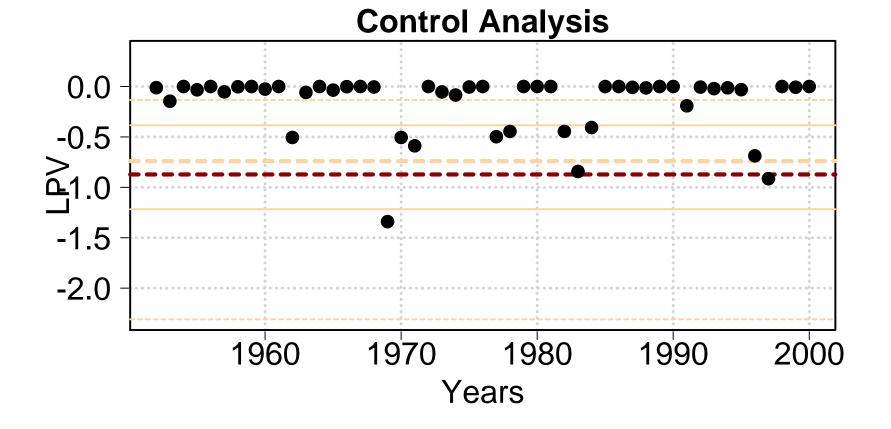


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Precipitation: CASE 2



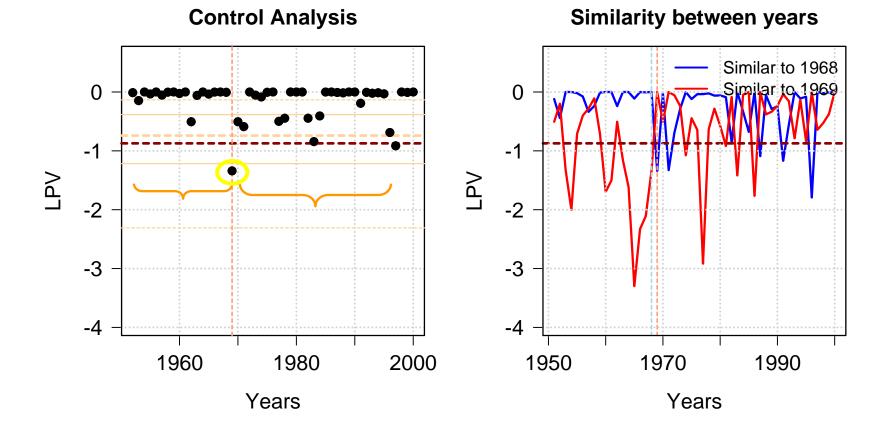


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Precipitation: CASE 2

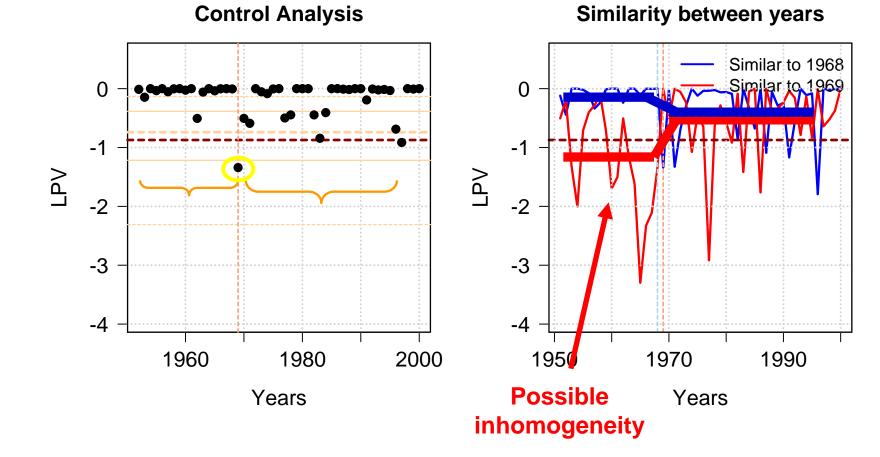








Precipitation: CASE 2



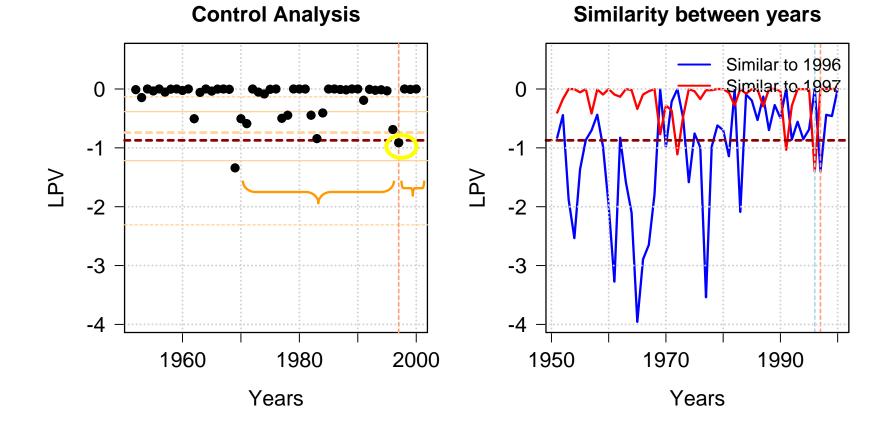


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Precipitation: CASE 2



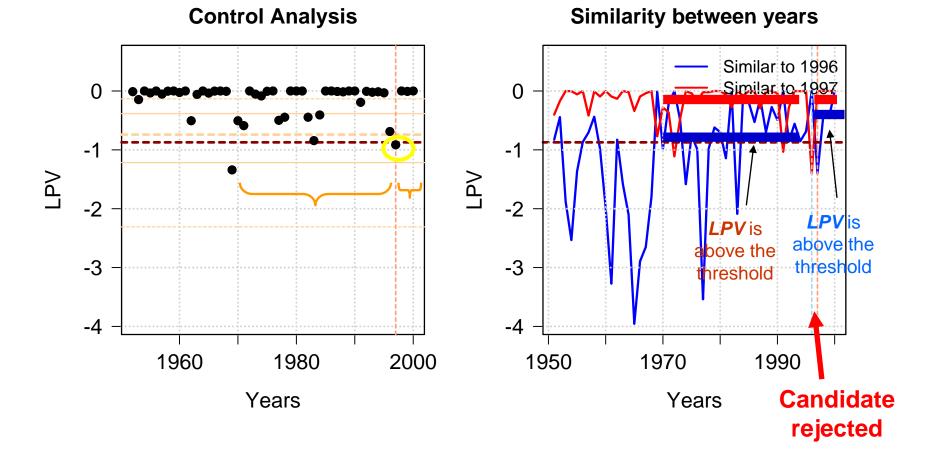


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Precipitation: CASE 2





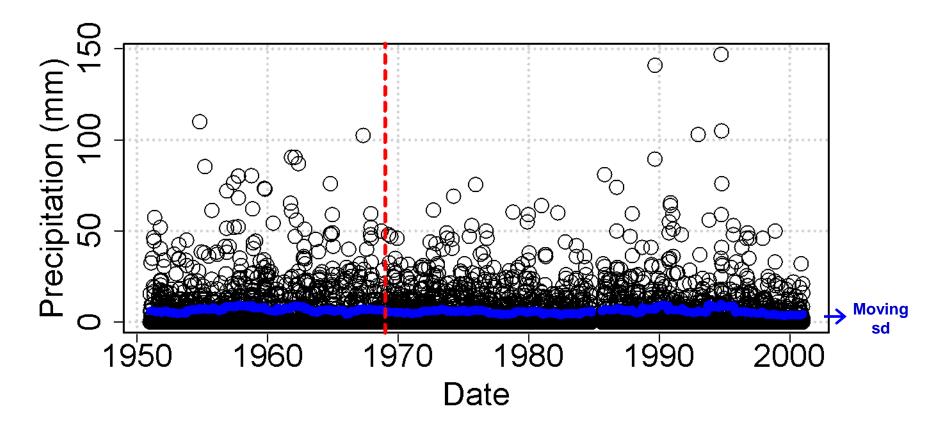
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Precipitation: CASE 2

Inhomogeneities in precipitation: final analysis





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- Selection of non-parametric test:
 - Kolmogorov-Smirnov: It is adequate for temperature and precipitation.
 - Other? Anderson-Darling: For precipitation is too much sensitive due to the high natural variability (takes a lot of extremes as "unusual year")



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 - Kolmogorov-Smirnov: It is adequate for temperature and precipitation.
 - Other? Anderson-Darling: For precipitation is too much sensitive due to the high natural variability (takes a lot of extremes as "unusual year")
- Analysis of control:
 - Introduction of the artificial inhomogeneity: Choice of parameters of the control (a=?, b =?). Our case, a = 3 for precipitation, b = 2 for temperature
 - Percentile of the LPV of reference. Level of confidence for a candidate? For precipitation can be important due to the high uncertainty.



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- Selection of non-parametric test:
 - Kolmogorov-Smirnov: It is adequate for temperature and precipitation.
 - Other? Anderson-Darling: For precipitation is too much sensitive due to the high natural variability (takes a lot of extremes as "unusual year")
- Analysis of control:
 - Introduction of the artificial inhomogeneity: Choice of parameters of the control (a=?, b =?). Our case, a = 3 for precipitation, b = 2 for temperature
 - Percentile of the LPV of reference. Level of confidence for a candidate? For precipitation can be important due to the high uncertainty.
- Inhomogeneity detection:
 - Inhomogeneity (on the left/right) must be greater than the reference jump
 - Size of the detection window is limited by the candidate neighbors.
 - Time-series with trends are a priori considered homogenious because increase is soft (no jumps), except if there is a long gap.



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INTRODUCTION

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About the methodology

- Non-parametric test of Kolmogorov-Smirnov can be used to detect inhomogeneity for both temperature and precipitation at daily scale.

- A control analysis is required introducing an artificial inhomogeneity ("reference jump") for comparing with the original time-series: LPV diagrams.

- For temperature, the recommended "reference jump" is 2°C and for precipitation is a multiplication factor of 3.



Conclusions





About the methodology

- Non-parametric test of Kolmogorov-Smirnov can be used to detect inhomogeneity for both temperature and precipitation at daily scale.

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About its aplication

- Temperature shows a reference LPV with less uncertainty than precipitation.
- Test can detect (sistematic) changes in ECDF, not only in the averages
- Some candidates of precipitation can be due to natural variability.
- A more extensive study of real cases could refine the methodology.







Thank you for your attention

and for getting up so early... ... as me.