Detecting inhomogeneities caused by methodological changes using MASH software

Beatrix Izsák¹, Olivér Szentes^{1,2}, Kinga Bokros^{1,2}

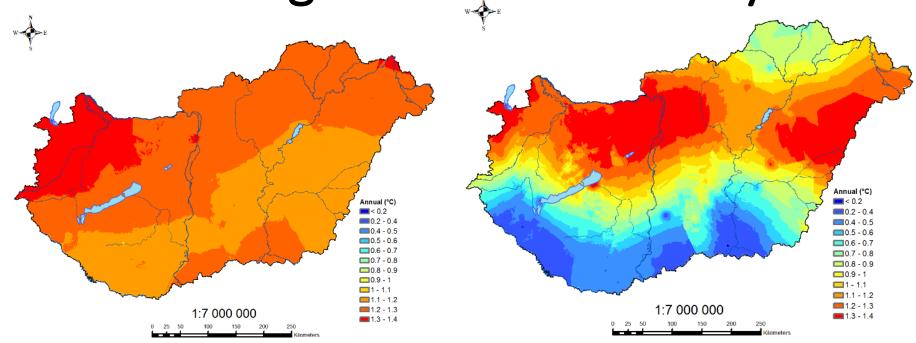
¹HungaroMet Hungarian Meteorological Service, Department of Climate
Research, Budapest, Hungary

²ELTE Eötvös Loránd University, Faculty of Science, Doctoral School of Earth
Sciences, Budapest, Hungary

Data Management Workshop 2025. Oslo



Is homogenization necessary?



Annual mean temperature, homogenized series: Estimation of change over the total period 1901-2018 (°C)

Annual mean temperature, raw series:

Estimation of change over the total period 1901-2018 (°C)

MASHv3.03

(Multiple Analysis of Series for Homogenization; *Szentimrey, T.*)

For homogenization, quality control and missing value completion of station daily data series

https://www.met.hu/en/rolunk/rendezvenyek/homogenization_and_inter
polation/software/

MASHv4.01

For homogenization in mean and standard deviation, quality control and missing value completion of station daily data series



Methodological changes

- Information about methodological changes is not always available.
 Nowadays, a large amount of paper-based measurements are being digitized in Hungary.
- In the case of temperature, there are 10-12 different methods alone. These newly recorded data must be verified.
- The MASH software also has a very good data check for daily data.
- Station relocation, instrument change and methodological change often occur simultaneously. Often, we cannot find an explanation for large inhomogeneities in the META data.
- In today's presentation, we will present cases where the inhomogeneity can be attributed to a methodological change. We show examples of the inhomogeneities detected by MASH that clearly result from methodological changes.

Statistical spatiotemporal modelling of monthly series in practice (Monthly series for a given month in a small region)

Relative Additive Model (normal distribution, e.g. temperature)

$$X_{j}(t) = \mu(t) + E_{j} + IH_{j}(t) + \varepsilon_{j}(t)$$
 $(j = 1, 2, ..., N; t = 1, 2, ..., n)$

Relative Multiplicative Model (e.g. precipitation)

$$X_{j}(t) = \mu(t) \cdot E_{j} \cdot IH_{j}(t) \cdot \exp(\varepsilon_{j}(t))$$

 μ : unknown climate change signal; E: spatial trend;

IH: inhomogeneity signal; ε : normal noise

Type of inhomogeneity IH(t) in general: 'step-like function'

Noise
$$\varepsilon(t) = [\varepsilon_1(t), ..., \varepsilon_N(t)]^T \in N(\mathbf{0}, \mathbf{C}) \ (t = 1, ..., n)$$
 are independent

Difference series constitution (Relative model)

Aim: to filter out the unknown climate signal $\mu(t)$.

 $X_{j}(t)$: candidate series; $X_{i}(t)$ $(i \neq j)$: reference series

Difference series for reference series $X_i(t)$ $(i \neq j)$:

$$Z_{j}(t) = X_{j}(t) - \sum_{i \neq j} \lambda_{ji} X_{i}(t) = IH_{j}(t) - \sum_{i \neq j} \lambda_{ji} IH_{i}(t) + \varepsilon_{Z_{j}}(t)$$

where sum of weighting factors: $\sum_{i \neq j} \lambda_{ji} = 1$

Difference series:
$$Z(t) = IH_Z(t) + \varepsilon_Z(t)$$
 $(t = 1,...,n),$

 $IH_Z(t)$: inhomogeneity with K break points, $T_1 < T_2 < < T_K$

$$\varepsilon_Z(t) \in N(E_Z, \sigma_Z^2)$$
 $(t = 1, ..., n)$ are independent

Optimal difference series constitution (MASH)

(interpolation between series)

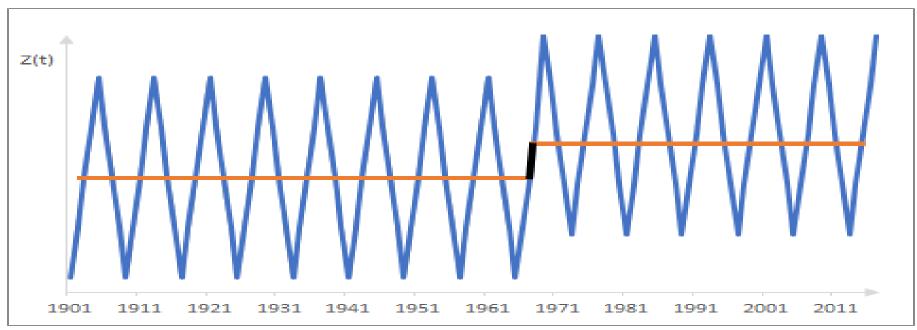
Minimization of the variance of noise $V(\varepsilon_{Z_j}) = V(Z_j)$.

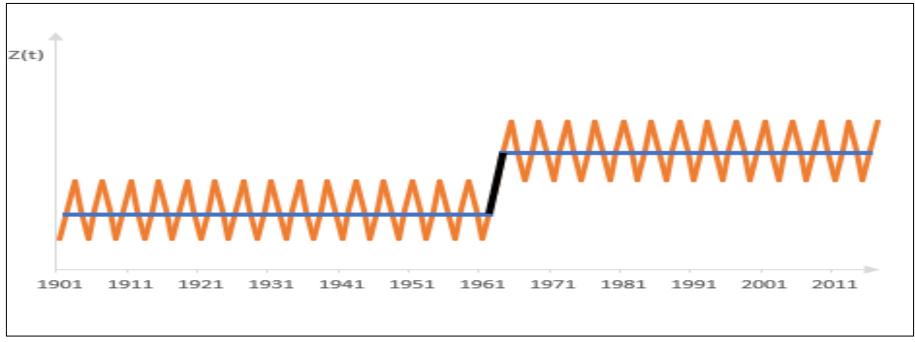
The optimal weighting factors λ_{ji} $(i \neq j)$ in vector form:

$$\lambda_{j} = \mathbf{C}_{j,ref}^{-1} \left(\mathbf{c}_{j,ref} + \frac{\left(1 - \mathbf{1}^{\mathrm{T}} \mathbf{C}_{j,ref}^{-1} \mathbf{c}_{j,ref} \right)}{\mathbf{1}^{\mathrm{T}} \mathbf{C}_{j,ref}^{-1} \mathbf{1}} \mathbf{1} \right)$$

 $\mathbf{c}_{j,ref}$: candidate-reference covariance vector,

 $\mathbf{C}_{i,ref}$: reference-reference covariance matrix





MASH: Multiple break points detection

- Break point (changepoint) detection for monthly series

 Examination (more) difference series to detect the break points

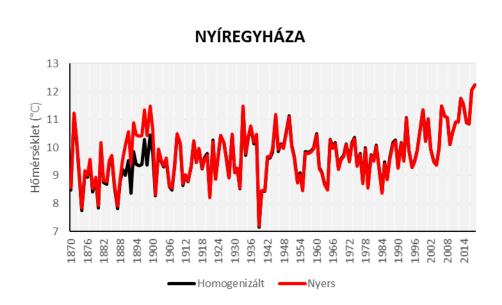
 and to attribute (separate) for the candidate series.
- Multiple break points detection based on **Test of Hypothesis**, confidence intervals for the break points, that make possible automatic use of metadata: MASH (*Szentimrey*)

Algorithm of MASH iteration procedure

- 1. To choose the candidate series
- 2. Series comparison: constitution of optimal difference series system
- 3. Break points detection for difference series: hypothesis testing, point estimation, confidence intervals
- 4. Estimation of shifts for difference series: point estimation, confidence intervals
- 5. Analysis of results: separation of break points and shifts for candidate series
- 6. Correction of candidate series if it is possible

Iteration of steps 1-6! Each series is examined many times!

Change of measurement times

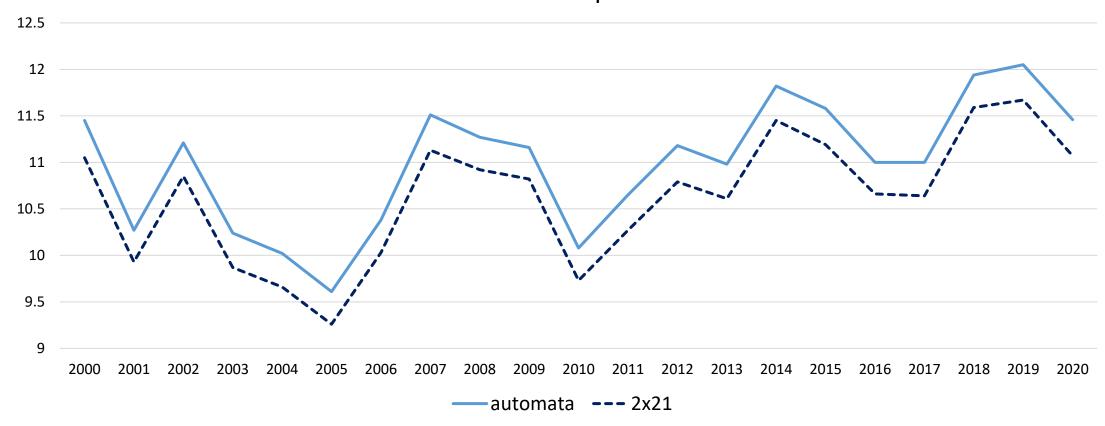


In Nyíregyháza, the change in measurement times causes inhomogeneity of the same magnitude as the detected climate change.

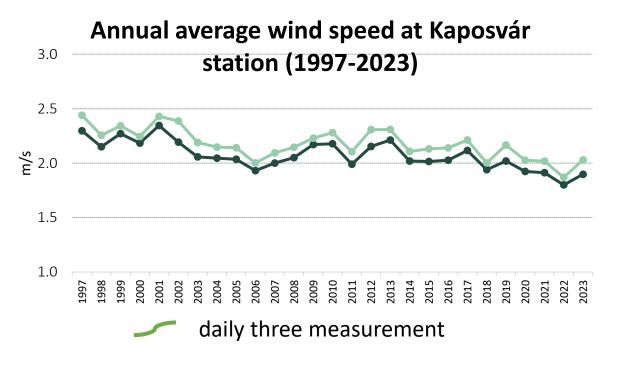
Annual temperature mean values calculated from raw homogenized data series, Nyíregyháza station. The graph clearly shows that from 1890 to 1901 the homogenized series is well below the raw data series, which is caused by the fact that the observations were made at different times. The morning measurement is 1 hour later, the evening measurement is 1 hour earlier.

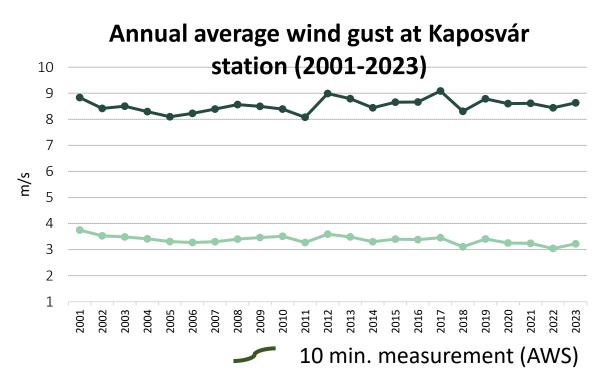
Automatic measurements and average of 06,12 and 2x21 h measurements (°C)

Annual mean temperature



Annual average wind and gusts calculated from three daily measurements and automatic measurements

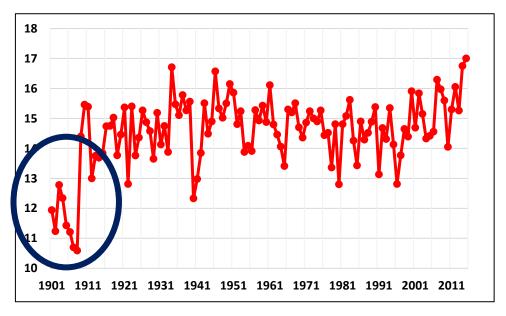




In the case of wind measurements, the methodological change always occurred together with the instrument change and relocation. Here we can only estimate how much of the detected inhomogeneity is due to the methodological change.

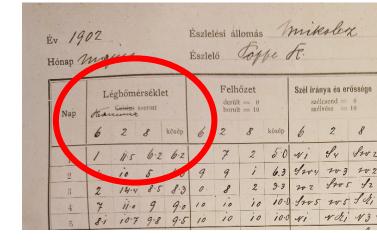


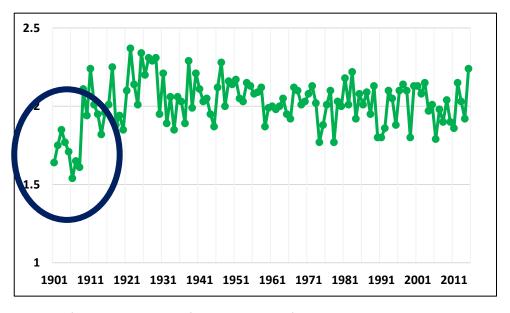
Relative Meteorológia feljegyzések **Evening** humidity measurement taken 3 hours later Mosonmagyaróvár 18 Meteorological measurements in June 1962, Mosonmagyaróvár (Source: HMS) Inhomogenity (%) 10 -18utc 4 2 0



Annual average series of monthly series $\overline{Y}(t)$ for characterization of annual mean (E).

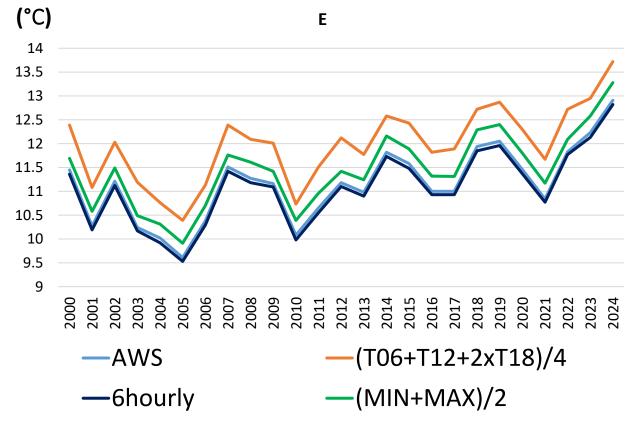
Inhomogeneity in 1901-1908, measured in Réaumur: 1 °C=0.8 Re





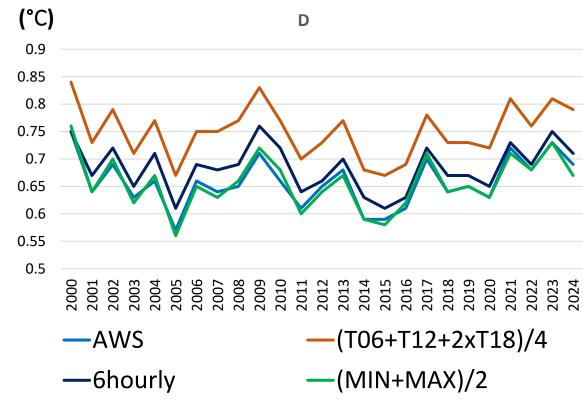
Series for characterization of annual average of the monthly standard deviations (D) based on series S(t).

Not only does the expected value change with the change in methodology, but so does the standard deviation.



Annual average temperature, 2000-2024

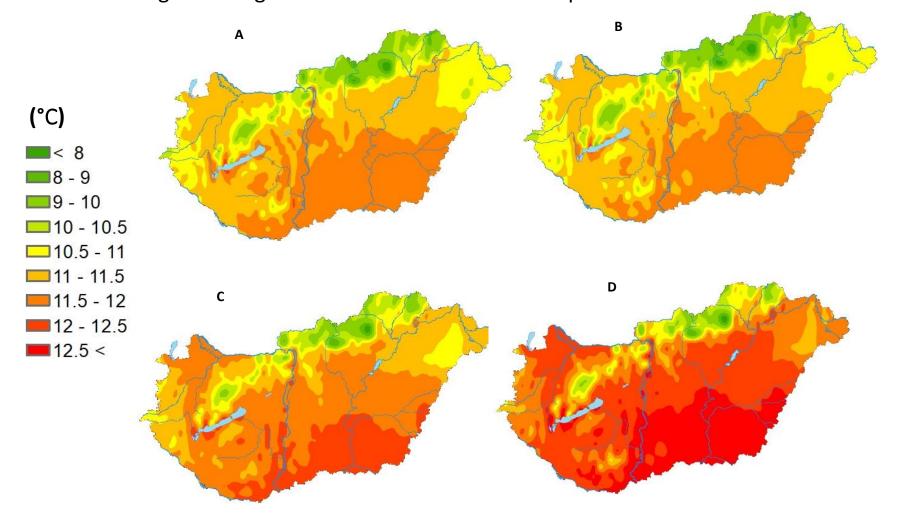




Spatial standard deviation of annual mean temperature (empirical standard deviation), national average, 2000–2024

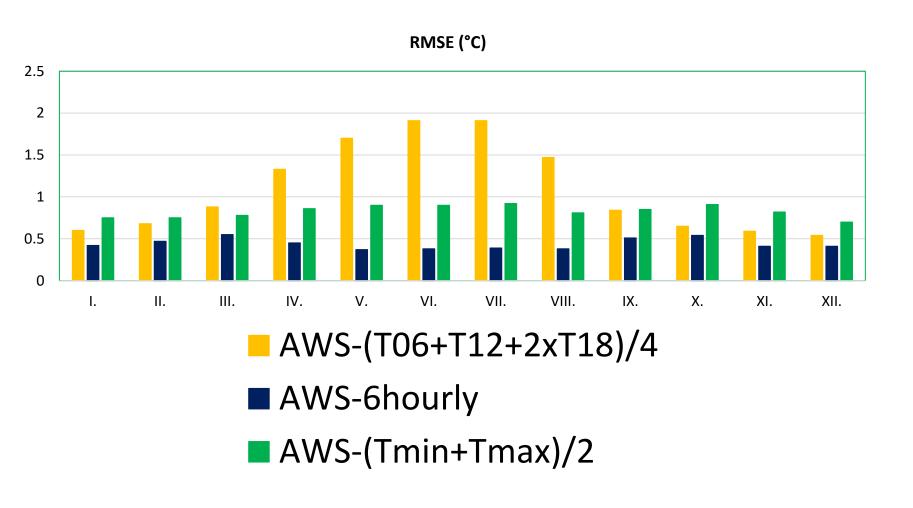
Today, with automatic measurements, we can simulate the change in measurement time in a simple way.

The impact of methodological changes is also different in terms of space.



Average annual temperature 2000-2025, A: automatic measurements, B: 4 measurements per day, C: (Tmin+Tmax)/2, D: evening measurements doubled

The methodological change has an annual cycle.



Software MASHv4.01 (Multiple Analysis of Series for Homogenization)

(**2023**, *T. Szentimrey*)

The MASH system is based on homogenization of monthly series derived from daily series. The procedures depend on the distribution of climate elements.

Quasi normal distribution (e.g. temperature)

Beside the monthly mean series another type of monthly series is also derived to estimate the inhomogeneity of standard deviation (D). These series are homogenized in standard deviation (D) by multiplicative model. The monthly mean series adjusted with the estimated inhomogeneity of standard deviation (D) are homogenized by additive model in mean (E).

Quasi lognormal distribution (e.g. precipitation)

Monthly mean or sum series are homogenized by multiplicative model.

Theorem (Problem of inhomogeneity of the standard deviation)

Daily data: Y(t) (t = 1,...,30), monthly mean: $\bar{Y} = \frac{1}{30} \sum_{t=1}^{30} Y(t)$

Monthly variable for examination of standard deviation (D): $S = \sqrt{\frac{1}{29} \sum_{t=2}^{30} (Y(t) - Y(t-1))^2}$

Daily data with inhomogeneity in mean (E) and standard deviation (D):

$$E(Y_{ih}(t)) = E(Y(t)) + \beta,$$

$$D(Y_{ih}(t)) = \alpha \cdot D(Y(t)) \qquad (t = 1, ..., 30)$$

The appropriate monthly variables: $\bar{Y}_{ih} = \frac{1}{30} \sum_{t=1}^{30} Y_{ih}(t)$,

$$S_{ih} = \sqrt{\frac{1}{29} \sum_{t=2}^{30} (Y_{ih}(t) - Y_{ih}(t-1))^2}$$

i, Then the monthly mean is also inhomogeneous in mean (E) and st. deviation (D):

$$E(\bar{Y}_{ih}) = E(\bar{Y}) + \beta$$
 and $D(\bar{Y}_{ih}) = \alpha \cdot D(\bar{Y})$

ii, Moreover variable S_{ih} can be used to estimate the inhomogeneity of st. deviation (D):

$$E(S_{ih}) = \alpha \cdot E(S)$$

Steps (Homogenization of monthly series S(t), $\bar{Y}(t)$):

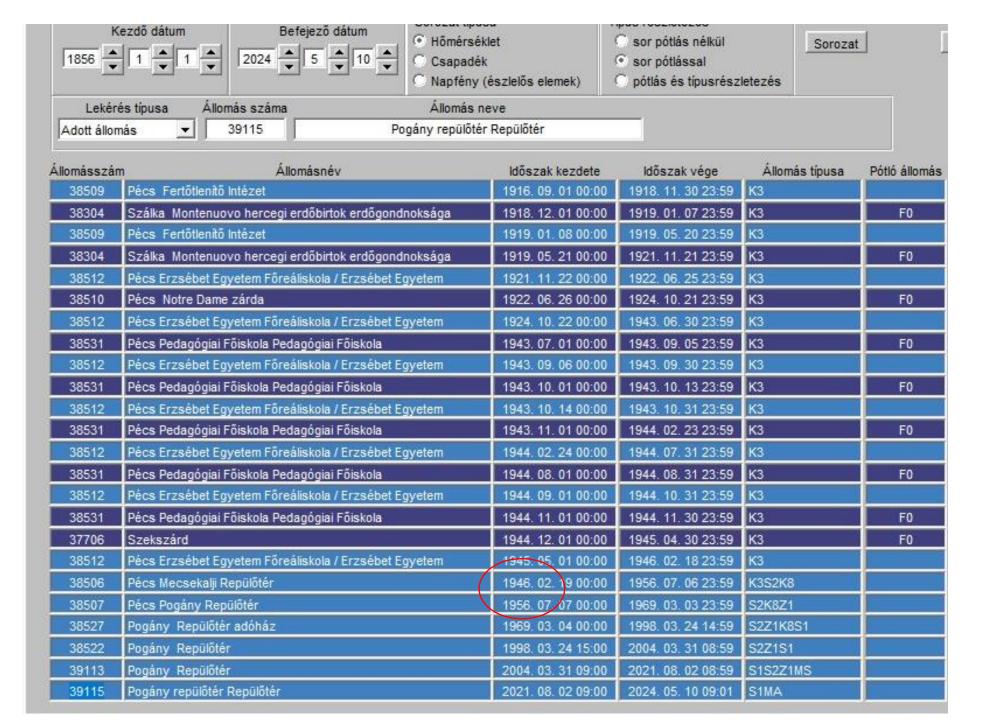
- Homogenization of series S(t) by multiplicative model: break points detection, estimation of inhomogeneity of standard deviation (D).
- Adjustment of standard deviation of series $\bar{Y}(t)$.
- Homogenization of adjusted series $\bar{Y}(t)$ by additive model: break points detection, estimation of the inhomogeneity of mean (E).
- Adjustment of mean of series $\bar{Y}(t)$.

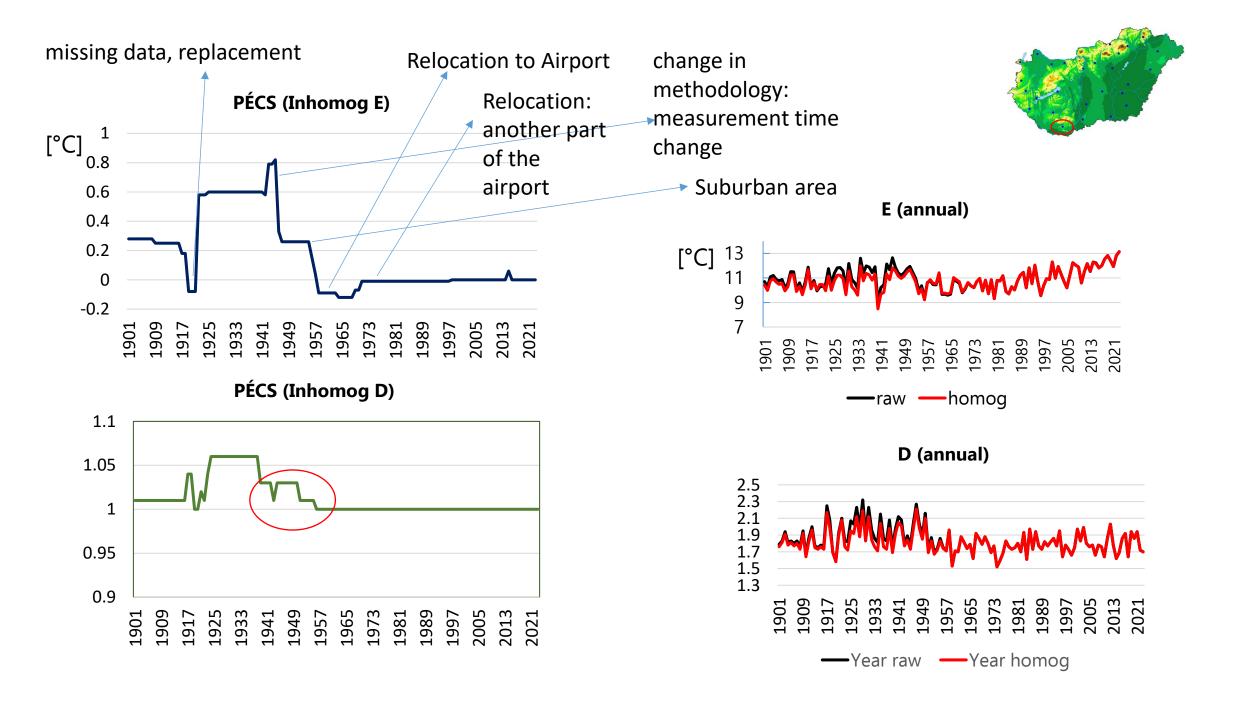
Assumption: homogeneity in higher order (>2) moments. This assumption is always right in case of normal distribution!

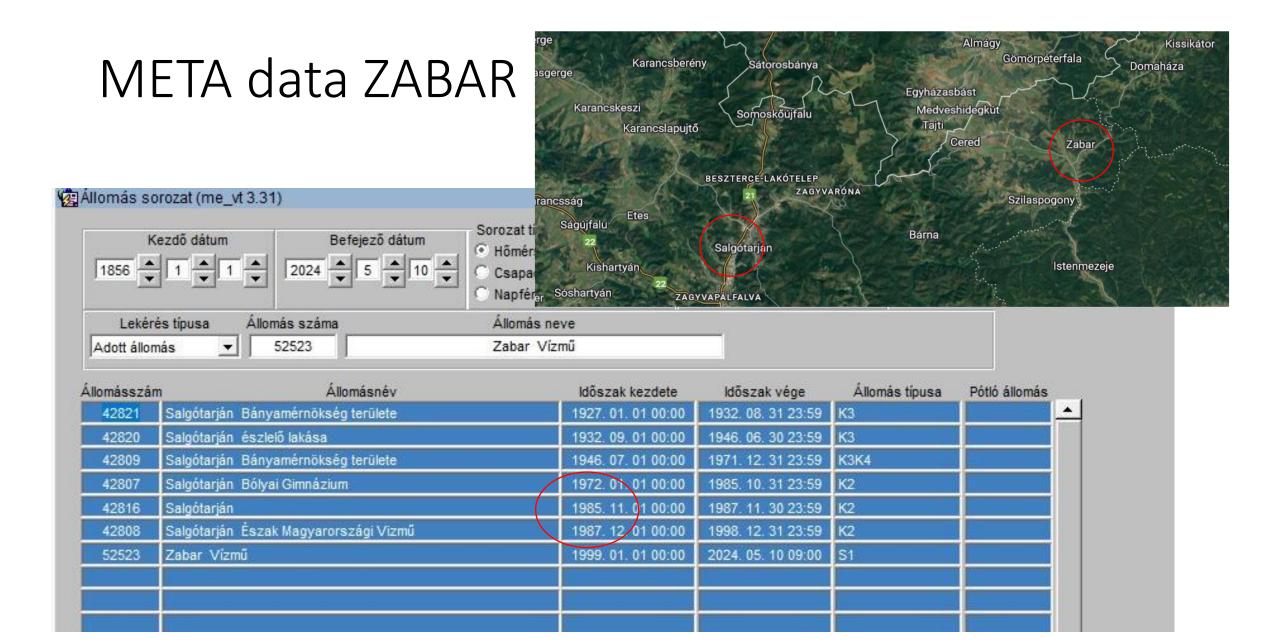
HOMOGENIZATION OF DAILY SERIES

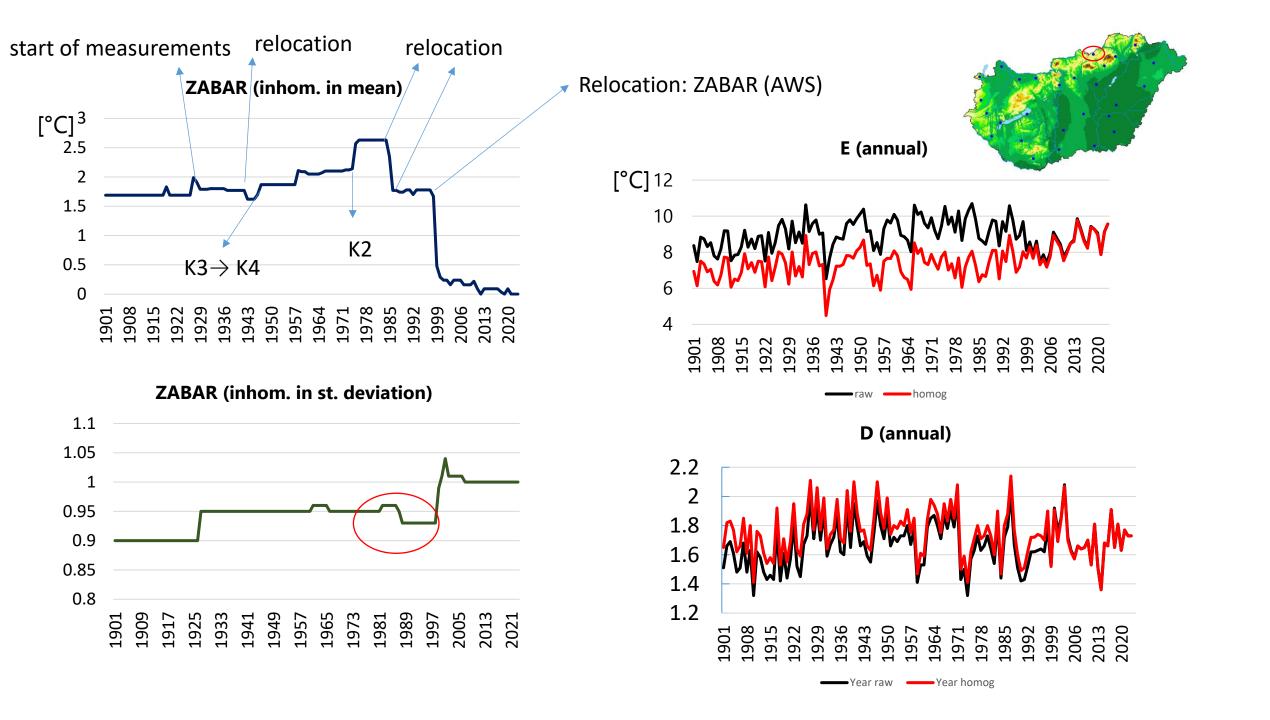
- Homogenization of mean (E) and standard deviation (D) on the basis of the monthly results.
- The used monthly information are the break points and the monthly adjustments of the mean (E) and standard deviation (D).
- If the daily data are normally distributed then there is no inhomogeneity in the higher order moments according to *Theorem 2* in the papers of *Szentimrey*, 2021, 2023.
- Szentimrey, T., 2021: Mathematical questions of homogenization and summary of MASH, Proceedings of the 10th Seminar for Homogenization and Quality Control in Climatological Databases and 5th Conference on Spatial Interpolation Techniques in Climatology and Meteorology (Ed. Lakatos M, Hoffmann L, Kircsi A, Szentimrey T), Budapest, Hungary, 2020, WCDMP-No. 86, pp. 4-17
- Szentimrey, T., 2023: Overview of mathematical background of homogenization, summary of method MASH and comments on benchmark validation, International Journal of Climatology, 43(13), 6314–6329 https://doi.org/10.1002/joc.8207

META data PÉCS









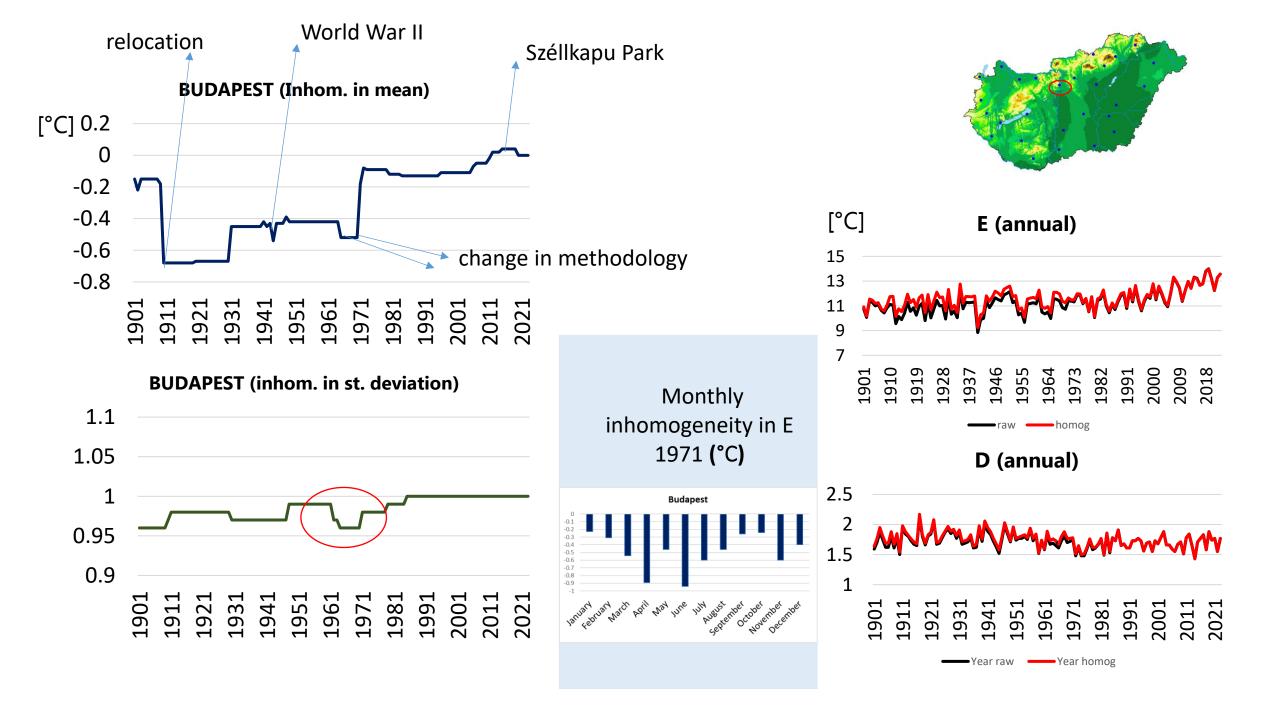




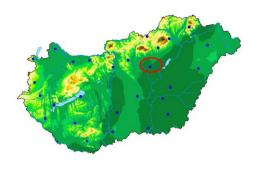
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Állomásszám	Állomásnév	ldőszak kezdete	ldőszak vége	Állomás típusa
44101	Budapest Budai vár Csillagvizsgáló	1856. 01. 01 00:00	1860. 12. 31 23:59	K3
44116	Budapest Víziváros Reáliskola (később:Toldy Ferenc Gimnázium	1861. 03. 14 00:00	1870. 12. 19 23:59	K3
44117	Budapest Várnegyed Hofhauser Casino	1870. 12. 20 00:00	1872. 09. 30 23:59	K3
44118	Budapest Krisztinaváros Novák villa - felső állomás	1872. 10. 01 00:00	1890. 12. 31 23:59	K3
44172	Budapest Vár alsó állomás	1891. 01. 01 00:00	1900. 01. 31 23:59	K3
44119	Budapest Víziváros	1900. 02. 01 00:00	1910. 02. 28 23:59	K3
44120	Budapest Országút KMI	1910. 03. 01 00:00	1985. 03. 31 23:59	K3K4
44120	Budapest Országút KMI	1910 03. 01 00:00	1960. 12. 31 23:59	K3
44120	Budapest Országút KMI	1961. 01. 0 00:00	1985. 03. 31 23:59	K4
44121	Budapest belterület OMSZ Torony	1985. 04. 01 00:00	2025. 10. 20 13:21	K4K8S2S1
44121	Budapest belterület OMSZ Torony	1985. 04. 01 00:00	1992. 03. 31 23:59	K4
44121	Budapest belterület OMSZ Torony	1992. 04. 01 00:00	1993. 08. 31 23:59	K8
44121	Budapest belterület OMSZ Torony	1992. 04. 01 00:00	1993. 08. 31 23:59	S2
44121	Budapest belterület OMSZ Torony	1993. 09. 01 00:00	2013. 04. 07 19:59	K4
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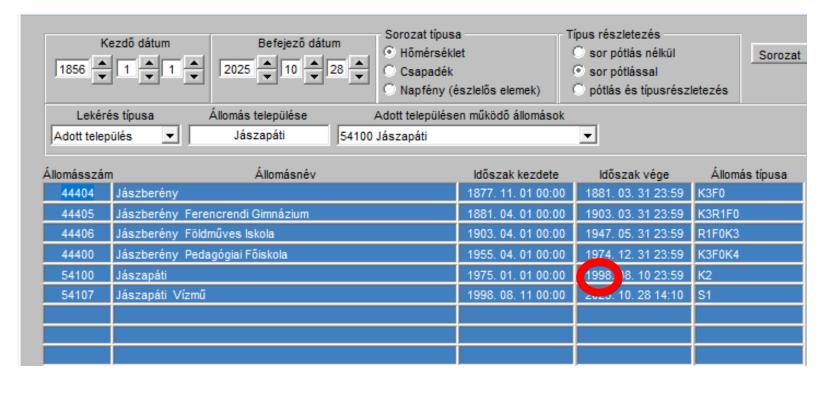
Széllkapu Park

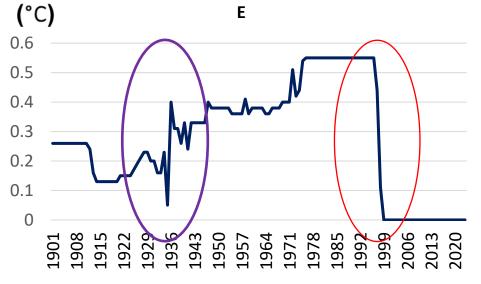
META data BUDAPEST

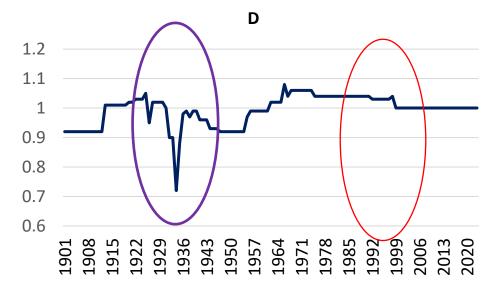


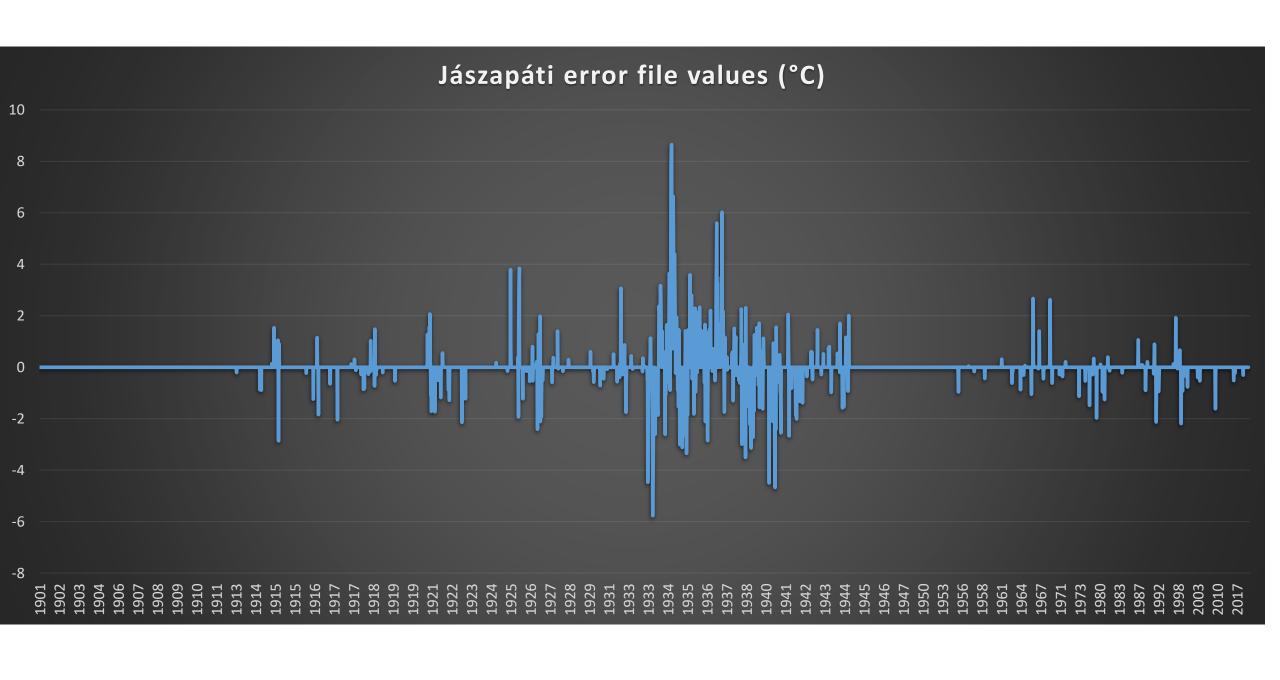
Jászapáti











Conclusion

- The inhomogeneities detected by the MASHv3.03 and MASHv4.01 software in the archive data can in many cases be explained by a change in methodology. Their magnitude and direction are consistent with the simulated data.
- In the case of temperature, the change in methodology can cause inhomogeneity of up to 2 degrees, while in the case of relative humidity, the largest detected and calculated inhomogeneity is approximately 16%.
- In the case of precipitation, the change in methodology did not cause much inhomogeneity in Hungary, given that precipitation was measured in the morning hours in all cases. (However, automatic precipitation gauges provide different precipitation data than the traditional method.)
- In the case of wind measurements, we cannot separate the methodological change from the other factors, because in all cases, relocation (including measurement height) and instrument replacement took place at the same time. With MASHv4.01 software (in the case of an additive model), the inhomogeneity

caused by the methodological change can also be detected in the second moment.

Thank you for your attention!

Save the date!

The **12th Seminar for Homogenization** and Quality Control in Climatological Databases and the 7th Interpolation Conference will be organized in **Budapest**, at the headquarters of the HungaroMet Hungarian Meteorological Service, and online, from **5 to 7 May 2026**.

Contact: seminar@met.hu





