

Weighted gravitational time series smoothing

*S. M. Agayan, Sh. R. Bogoutdinov,
M. N. Dobrovolsky, and A. I. Kagan*

Geophysical Center of Russian Academy of Science,
Moscow, Russia

Abstract. This article continues the series of papers by the authors on the new universal DMA-smoothing of time series, originally intended for the analysis of geophysical time series obtained in the framework of discrete mathematical analysis (DMA), developed by GC RAS. We formulated the general concept of weighted DMA-smoothing, constructed and analyzed one of its variants.

Introduction

Time series are the most effective way to represent dynamic data, the results of various kinds of monitoring (physical, geophysical, environmental et al.). Smoothing of time series is one of the most powerful tools to study them, because the smoothed time series often are simpler, more amenable to simple analysis and serve as an effective approximation of the original time series.

To date, the most known methods of smoothing are regression analysis, finite element method and splines, Fourier and wavelet smoothing and their generalizations based on expansions in orthogonal systems of functions and time-frequency analysis.

The proposed method expands the list of a new type of smoothing, able to operate successfully in the analysis of geophysical time series. In the article it was applied to real data from the Borok geomagnetic observatory of the INTERMAGNET network.

Main Concept of DMA-Smoothing

Let $TS[a, b]$ be a space of time series on a discrete segment $[a, b]$ with nodes $t_i = a + (i-1)h$, $i = 1, \dots, n$, $h = (b - a)/(n - 1)$. The elements of $TS[a, b]$ are

denoted by x, y, z, \dots . If $z \in \text{TS}[a, b]$, then $z_i = z(t_i)$ and the correspondence $z \sim (z_i)|_1^n$ turns $\text{TS}[a, b]$ into n -dimensional Euklidian space $R^n = R^n(z_1, \dots, z_n)$.

We considered the task of constructing the smoothing $\text{Sm } y \stackrel{\text{def}}{=} x \in \text{TS}[a, b]$ for the given time series $y \in \text{TS}[a, b]$. Its solution is a consistent formalization of the following conjunction:

$$\begin{aligned} \text{"}x - \text{smoothing } y\text{"} \equiv & (x - \text{approximation } y) \wedge \\ & (x - \text{smoothed time series}) \end{aligned} \tag{1}$$

Let us denote by z a variable series in $\text{TS}[a, b]$. The approximation in (1) is formalized by the quadratic residual functional $\text{Sc}(z|y) = \|z - y\|^2$, and the smoothness – by the non-negative quadratic form $\text{CG}(z)$, which we call the smoothness residual. It is based on a variant of the discrete continuity. The residual $\text{CG}(z)$ is a measure of the deviation from this property for the time series z on the interval $[a, b]$. The less are the values $\text{Sc}(z|y)$ and $\text{CG}(z)$, the more reason we have to consider z the smoothing y . Consequently, we need a meaningful simultaneous minimization of the functionals $\text{Sc}(z|y)$ and $\text{CG}(z)$. To do this, we considered their nontrivial convex combination

$$\text{Sm}_\lambda(z|y) = \lambda \text{CG}(z) + (1 - \lambda) \text{Sc}(z|y), \lambda \in (0, 1)$$

and a search of the smoothing $x = \text{Sm} y$ for y is reduced to a minimization on $\text{TS}[a, b]$ at a suitable $\lambda \in (0, 1)$ of the functional $\text{Sm}_\lambda(z|y)$, i.e. to the solution of the n -dimensional linear system

$$x = \text{Sm}_\lambda y \Leftrightarrow \text{Grad Sm}_\lambda(x|y) = 0$$

Thus we suggest the whole family $\{\text{Sm}_\lambda y, \lambda \in (0, 1)\}$ as the smoothing $\text{Sm} y$. Let us find the gradient $\text{Grad Sm}_\lambda(x|y)$ in its clear form. For that first we introduce the non-negative on $\text{TS}[a, b]$ operator G , which generates the gravitational smoothness residual $\text{CG}(z) = (Gz, z)$ and then we transform $\text{Sm}_\lambda(z|y)$:

$$\text{Sm}_\lambda(z|y) = \lambda(Gz, z) + (1 - \lambda)\|z - y\|^2 =$$

$$\lambda(Gz, z) + (1 - \lambda)(z - y, z - y) =$$

$$\lambda(Gz, z) + (1 - \lambda)(z, z) - 2(1 - \lambda)(z, y) + (1 - \lambda)\|y\|^2 =$$

$$((\lambda G + (1 - \lambda)E)z, z) - 2(1 - \lambda)(z, y) + (1 - \lambda)\|y\|^2$$

Hence E is a unit operator on $\text{TS}[a, b]$. Consequently, the minimization $\text{Sm}_\lambda(z|y)$ is equal to the

functional minimization

$$\widetilde{\text{Sm}}_{\lambda}(z|y) = \frac{1}{2}((\lambda G + (1 - \lambda)E)z, z) - ((1 - \lambda)z, y)$$

For it the gradient in the point z is expressed through G like this [*Pshenichny and Danilin, 1975*]:

$$\text{Grad } \widetilde{\text{Sm}}_{\lambda}(z|y) = (\lambda G + (1 - \lambda)E)z - (1 - \lambda)y$$

Thus,

$$x = \text{Sm}_{\lambda}y \Leftrightarrow (\lambda G + (1 - \lambda)E)x = (1 - \lambda)y$$

If $\lambda \in (0, 1)$, then the operator $\lambda G + (1 - \lambda)E$ is positive on $\text{TS}[a, b]$, and therefore reversible. Consequently, the smoothing $x = \text{Sm}_{\lambda}y$ always exists, is uniquely defined for any $y \in \text{TS}[a, b]$ and represents the result of applying the operator $(\lambda G + (1 - \lambda)E)^{-1}$, therefore

$$\text{Sm}_{\lambda} = (1 - \lambda)(\lambda G + (1 - \lambda)E)^{-1}$$

Let us consider an important variant of the residual $\text{CG}(z)$.

Gravitational Residual of Smoothness

Let us briefly (according to the module [*Agayan et al., 2010; Gvishiani et al., 2011*]) consequently, from top

to bottom recall the construction of the main, but not the only one residual of the discrete smoothness $CG(z)$: the gravitational smoothness residual $CGr(z)$:

$$CGr(z) = \sum_{s=0}^{n-1} \omega_s CGr^s(z)$$

where ω_s is the non-negative weights (parameters of smoothness). The residual $CGr(z)$ was generated by the operator of gravitational smoothness Gr :

$$CGr(z) = (Gr z, z), \quad Gr = \sum_{s=0}^{n-1} \omega_s Gr^s$$

where

$$Gr^s = D^{s*} Gr[a, b^s] D^s$$

and D^s is the s -dimensional discrete differentiation operator: if z is the time series on the interval $[a, b]$, then its s -dimensional derivative $z^s = D^s z$ is time series on the interval $[a, b^s]$, $b^s = b - sh$ and in each node $t_i \in [a, b^s]$

$$z^s(t_i) = \frac{\sum_{l=0}^s (-1)^{s-l} C_s^l z_{i+l}}{h^s}$$

The operator $\text{Gr}[a, b^s]$ is called the operator of gravitation continuity residual on the interval $[a, b^s]$ and is defined by the proximity measure $\delta_{t_i}^s(t_j)$ in accordance with the following formula

$$\text{Gr}[a, b^s] = (A(\delta^s) - E)^*(A(\delta^s) - E)$$

$$A(\delta^s) = (a_{ij}^s)$$

$$a_{ij}^s = \frac{\delta_i^s(j)}{\sum_{j=1}^{n-s} (\delta_i^s(j))}$$

One of the following constructions was used as the proximity measure:

$$\delta_i^s(j) = \delta_{t_i}^s(t_j) =$$

$$\left(1 - \frac{|t_j - t_i|}{\max(|t_i - a|, |t_i - b_s|) + h} \right)^p$$

as shown in Figure 1, or

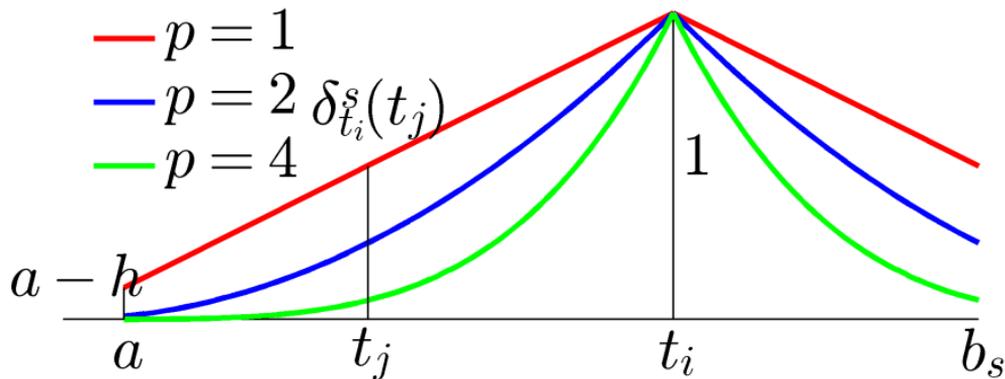


Figure 1. Global proximity measure.

$$\delta_i^s(j) = \delta_i^{s,r}(j) = \delta_{t_i}^{s,r}(t_j) = \begin{cases} \left(1 - \frac{|t_j - t_i|}{r}\right)^p & \text{if } |t_j - t_i| < r, \\ 0 & \text{if } |t_j - t_i| \geq r. \end{cases}$$

for a certain $r < b_s - a$, as shown in Figure 2

Parameters of Gravitational Smoothness

Thus there are three free parameters of gravitational smoothing: the coupling multiplier $\lambda \in [0, 1]$, the system w of non-negative weights w_s , $s = 0, \dots, n-1$ and

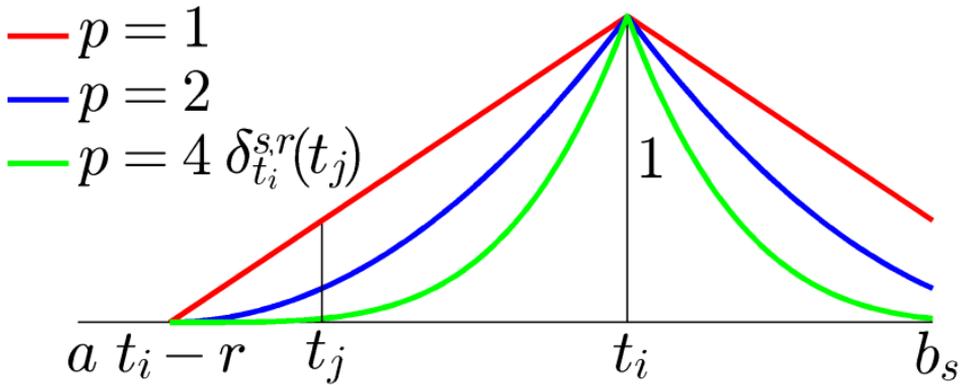


Figure 2. Local proximity measure.

the system δ of proximity measures δ^s , $s = 0, \dots, n-1$, which depends on the internal parameter $p \geq 0$.

These parameters give the gravitational smoothing greater flexibility and expressiveness. The study allows some qualitative conclusions about their relationship to be made:

1. the increment λ at fixed w and δ leads to increasing smoothness and reduces the scanning (Figure 3).
2. the increment p at fixed λ and w leads to improving scanning with the preservation of smoothness (Figure 4).
3. the influence of weights w at fixed λ and p on gravitational smoothness is more complicated and

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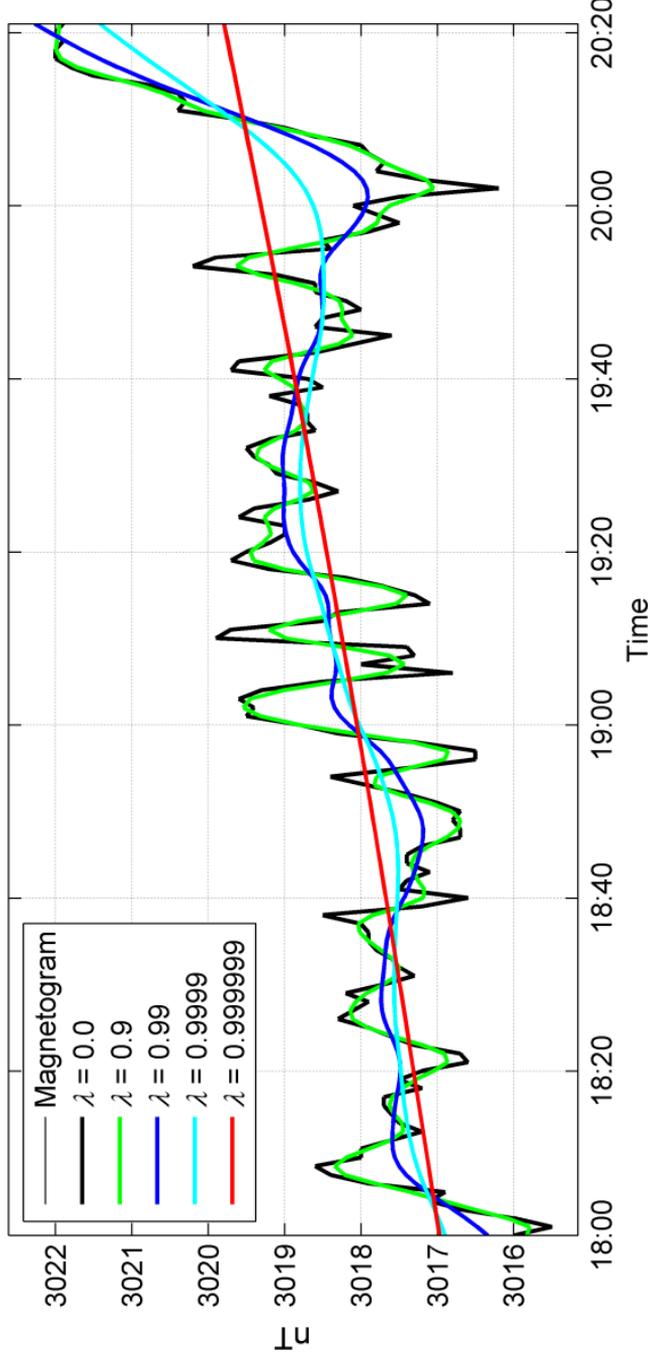


Figure 3. Dependence on λ at fixed w and δ .

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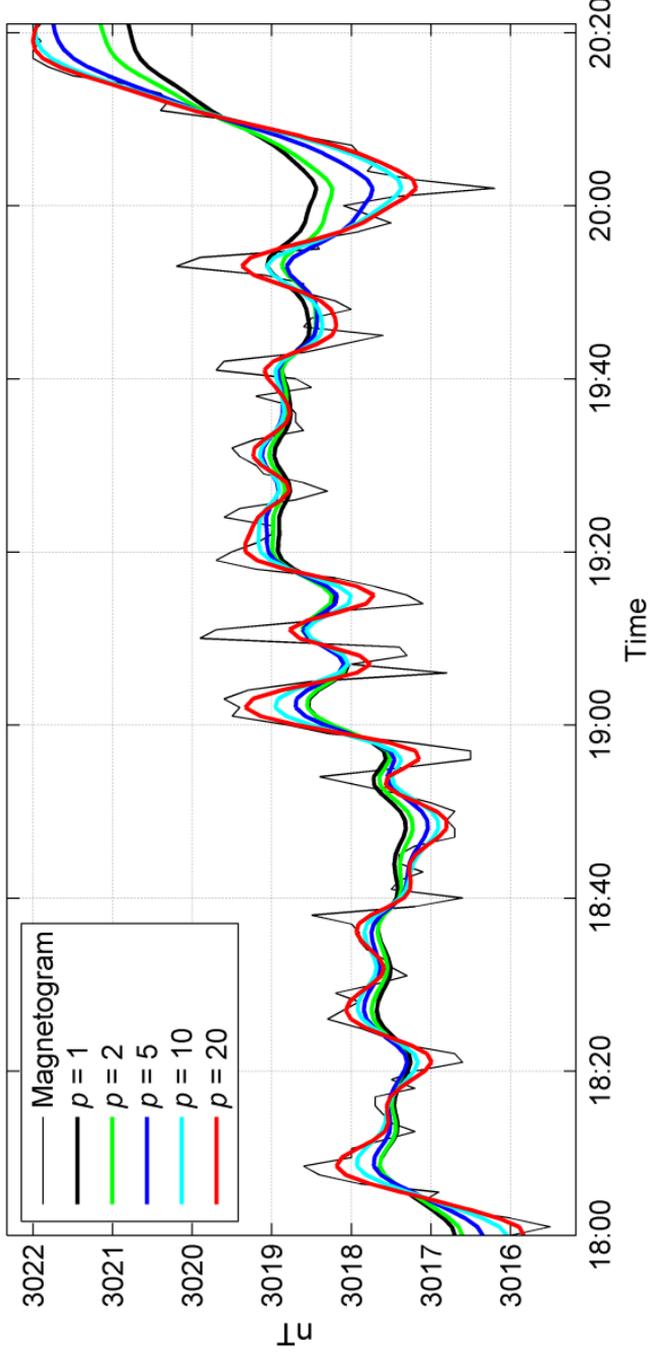


Figure 4. Dependence on p at fixed λ and δ .

depends on their relationship. In Figure 5 the simplest dependence on basic weights is shown.

4. our experience with the algorithm allowed the most appropriate values of the free parameters to be identified: $\lambda \in [0.9, 0.999]$, $p \in [2, 10]$, $w = \{w_0 = 0, w_{1-4} = 1, w_s = 0, s > 4\}$ (Figure 6).

Weighted DMA-Smoothing. Main Concept

Let us assume that each node t_i in the interval $[a, b]$ has its non-trivial weight $v_i \geq 0$. Let us denote by \mathbf{V} their total population $\mathbf{V} = \{v_i|_1^n\}$. This raises the question of the impact of \mathbf{V} on the DMA-smoothing, which is the subject of this paragraph. The functional

$$\text{Sm}_\lambda(x|y) = \lambda \text{CG}(x) + (1 - \lambda) \text{Sc}(x|y)$$

is responsible for the smoothing, so the impact \mathbf{V} occurs separately: through the smoothness residual $\text{CG}(x)$, the scanning functional $\text{Sc}(x|y)$ and the connecting multiplier λ .

In the future we shall also denote by \mathbf{V} the diagonal operator with weights v_i . Further, the smoothing

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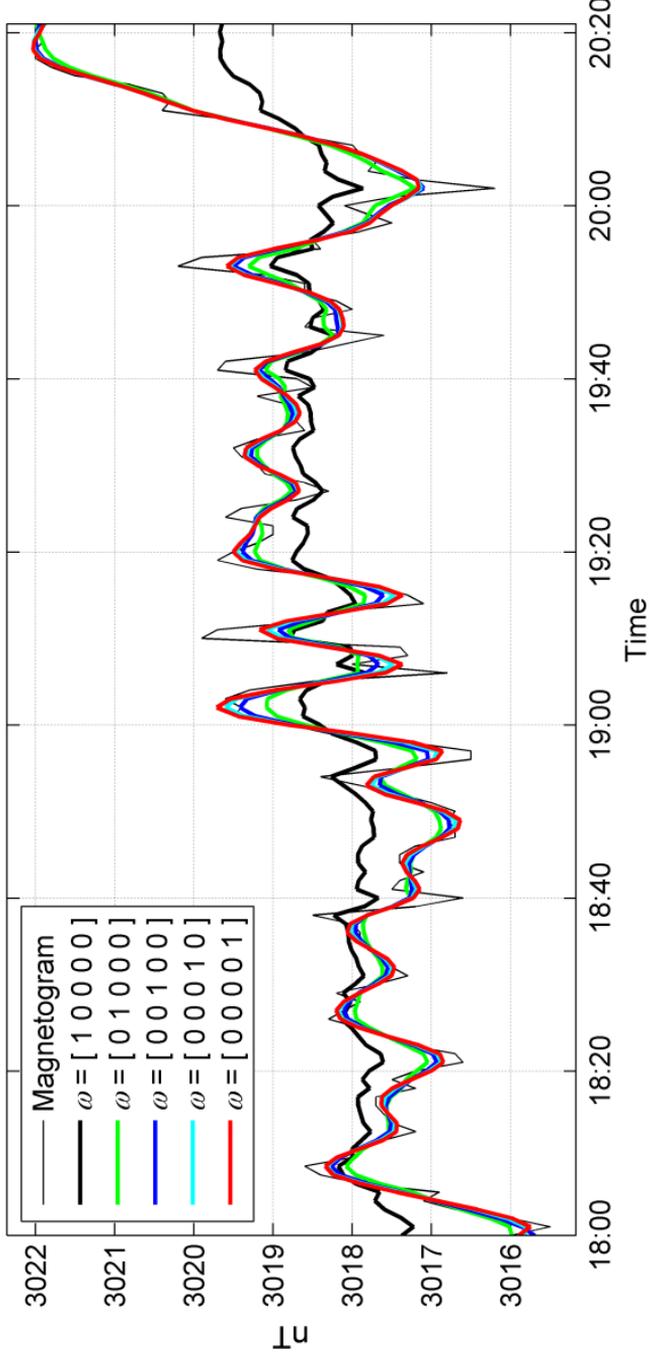


Figure 5. Dependence on w at fixed λ and p .

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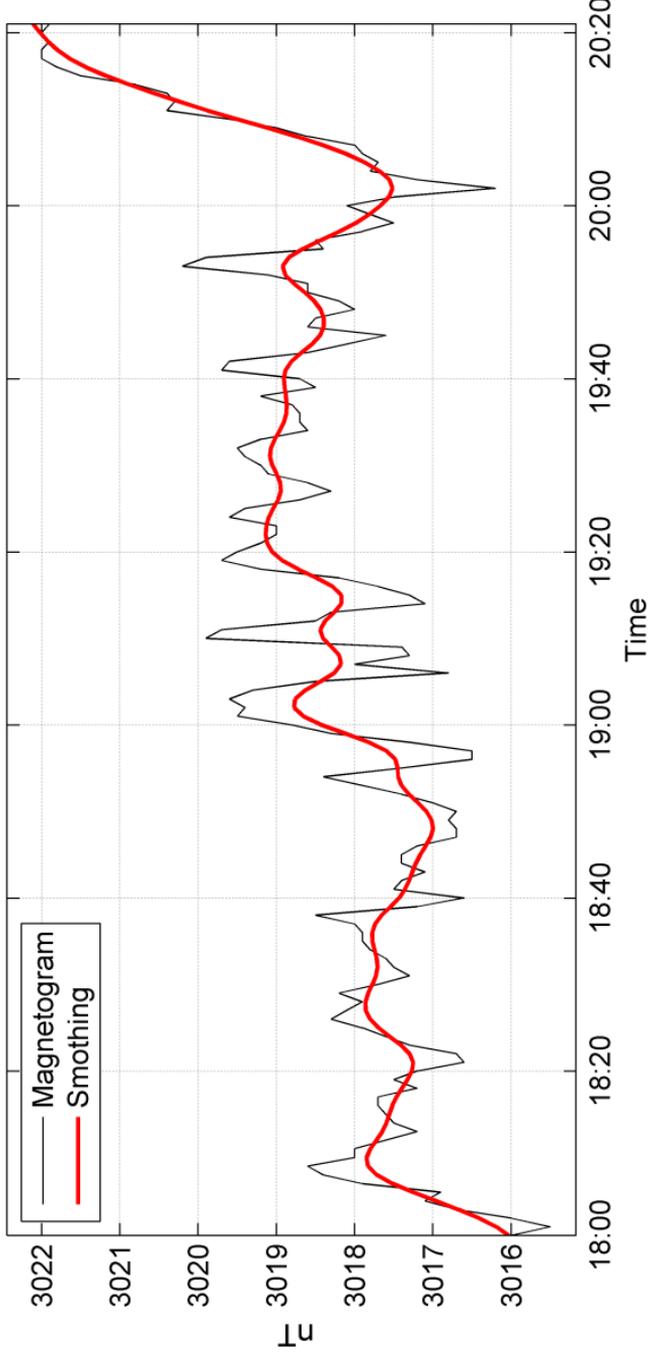


Figure 6. $\lambda = 0.95$, $p = 4$, $w = \{w_0 = 0, w_{1-4} = 1, w_s = 0, s > 4\}$.

elements, into which the weights \mathbf{V} have (or have not) penetrated, will be respectively denoted by $CG_V(x)$, $Sc_V(x|y)$, λ_V ($CG(x)$, $Sc(x|y)$, λ). Relationships between the smoothing Sm and the weights \mathbf{V} in a general case can develop according to one of the following eight scenarios:

$$Sm_\lambda(x|y) = \lambda CG(x) + (1 - \lambda) Sc(x|y)$$

$$VSm_\lambda(x|y) = \lambda CG(x) + (1 - \lambda) Sc_V(x|y)$$

$$SmV_\lambda(x|y) = \lambda CG_V(x) + (1 - \lambda) Sc(x|y)$$

$$VSmV_\lambda(x|y) = \lambda CG_V(x) + (1 - \lambda) Sc_V(x|y)$$

$$Sm_{\lambda_V}(x|y) = \lambda_V CG(x) + (1 - \lambda_V) Sc(x|y) \quad (2)$$

$$VSm_{\lambda_V}(x|y) = \lambda_V CG(x) + (1 - \lambda_V) Sc_V(x|y)$$

$$\text{SmV}_{\lambda_V}(x|y) = \lambda_V \text{CG}_V(x) + (1 - \lambda_V) \text{Sc}(x|y)$$

$$\text{VSmV}_{\lambda_V}(x|y) = \lambda_V \text{CG}_V(x) + (1 - \lambda_V) \text{Sc}_V(x|y)$$

The interest in the structures (2) is explained by the following circumstances: in general stochastic setting smoothing of time series can be considered as a possible option of ideal course of the process, which is described by the given time series. If the disturbances of process in time are expressed by weights, the weighted smoothing gives an idea of the impact of these disturbances on the process.

The simplest is the impact of \mathbf{V} on the scanning $\text{Sc}(x|y)$: it is natural to assume that

$$\text{Sc}_V(x|y) = \sum_{i=1}^n v_i (x_i - y_i)^2$$

If the weights v_i are positive, then the operator \mathbf{V} is reversible and the smoothing $x = \text{Smy}$ (2) looks like this

$$\text{Sm}_{\lambda} y = (1 - \lambda)(\lambda \mathbf{G} + (1 - \lambda)\mathbf{E})^{-1} y$$

$$VSm_{\lambda} y = (1 - \lambda)(\lambda G + (1 - \lambda)E)^{-1} \mathbf{V}y$$

$$SmV_{\lambda} y = (1 - \lambda)(\lambda G_V + (1 - \lambda)E)^{-1} y$$

$$VSmV_{\lambda} y = (1 - \lambda)(\lambda G_V + (1 - \lambda)E)^{-1} \mathbf{V}y$$

$$Sm_{\lambda_V} y = (1 - \lambda_V)(\lambda_V G + (1 - \lambda_V)E)^{-1} y \quad (3)$$

$$VSm_{\lambda_V} y = (1 - \lambda_V)(\lambda_V G + (1 - \lambda_V)E)^{-1} \mathbf{V}y$$

$$SmV_{\lambda_V} y = (1 - \lambda_V)(\lambda_V G_V + (1 - \lambda_V)E)^{-1} y$$

$$VSmV_{\lambda_V} y = (1 - \lambda_V)(\lambda_V G_V + (1 - \lambda_V)E)^{-1} \mathbf{V}y$$

To summarize, there are seven non-trivial scenarios of varying degrees of penetration depth of the weights \mathbf{V} of nodes in the DMA-smoothing Sm . The simplest is the second scenario VSm_{λ} ((2) and (3)) in which the weights of nodes penetrate the scanning only.

We conclude this work with the analysis of such smoothing in the gravitational case.

Weighted Gravitational Smoothing VSm_{λ}

In all the smoothings in Figure 7–Figure 9 the nodes of the main record y were supposed to be equal, having a

unit weight. Let us break this equality three times by giving

- in the first case to the points on the interval [18:30–18:50] the weight 10.0;
- in the second case to the points on the interval [19:50–20:10] the weight 0.1;
- in the third unifying case to the points on the interval [18:30–18:50] the weight 10.0, to the points on the interval [19:50–20:10] the weight 0.1.

In each case at the same parameters λ , w and p we constructed the smoothings $Sm_\lambda y$ and $VSm_\lambda y$. In Figure 7–Figure 9 the smoothing $Sm_\lambda y$ is shown in blue and outside the intervals [18:30–18:50] and [19:50–20:10] is concealed by the smoothing $VSm_\lambda y$, shown in red. Our study has shown that the situation in Figure 7–Figure 9 typical for the correlation of $Sm_\lambda y$ and $VSm_\lambda y$ and allows the following conclusion to be made: under the same conditions λ , w , p the smoothings $Sm_\lambda y$ and $VSm_\lambda y$ have the same smoothness, at that the construction $VSm_\lambda y$ is clearly dependent on the weights V : the higher is the weight v_i in the node t_i , the more carefully the smoothing $VSm_\lambda y$ scans the record y in the vicinity t_i .

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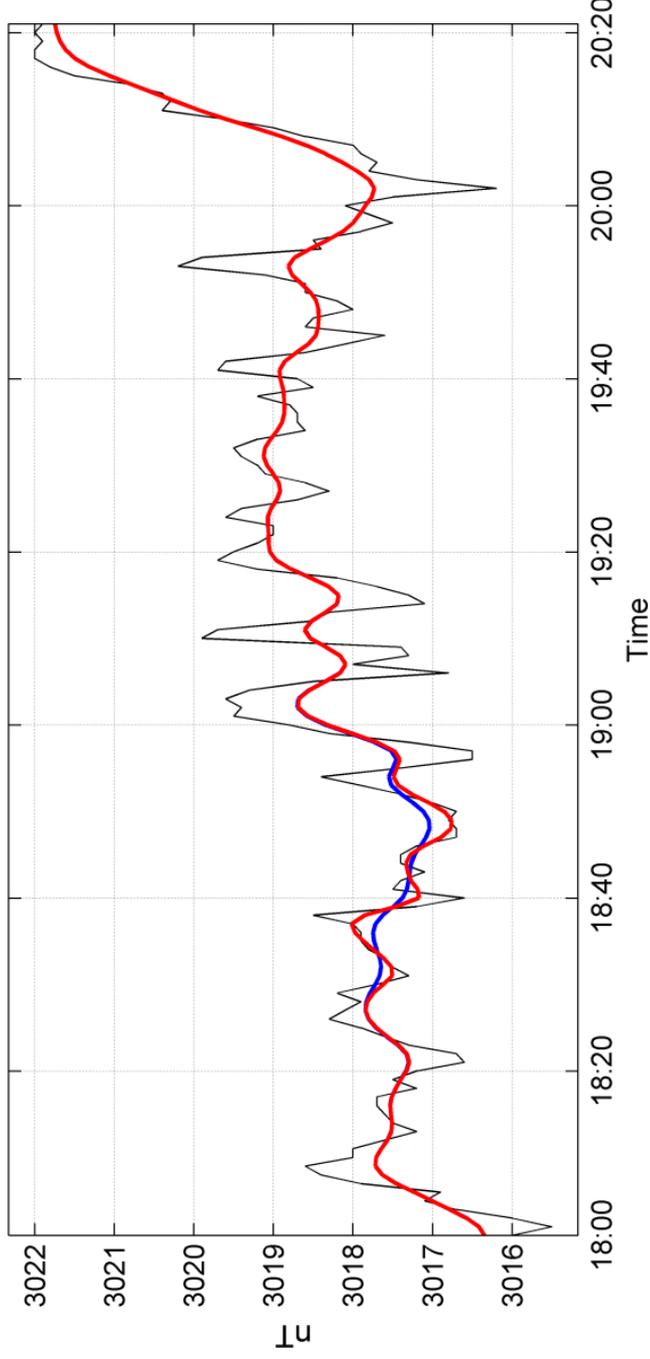


Figure 7. The points' weights in the period 18:30–18:50 are equal to 10.0.

Observatory BOX, Component Y (14-Nov-2008)

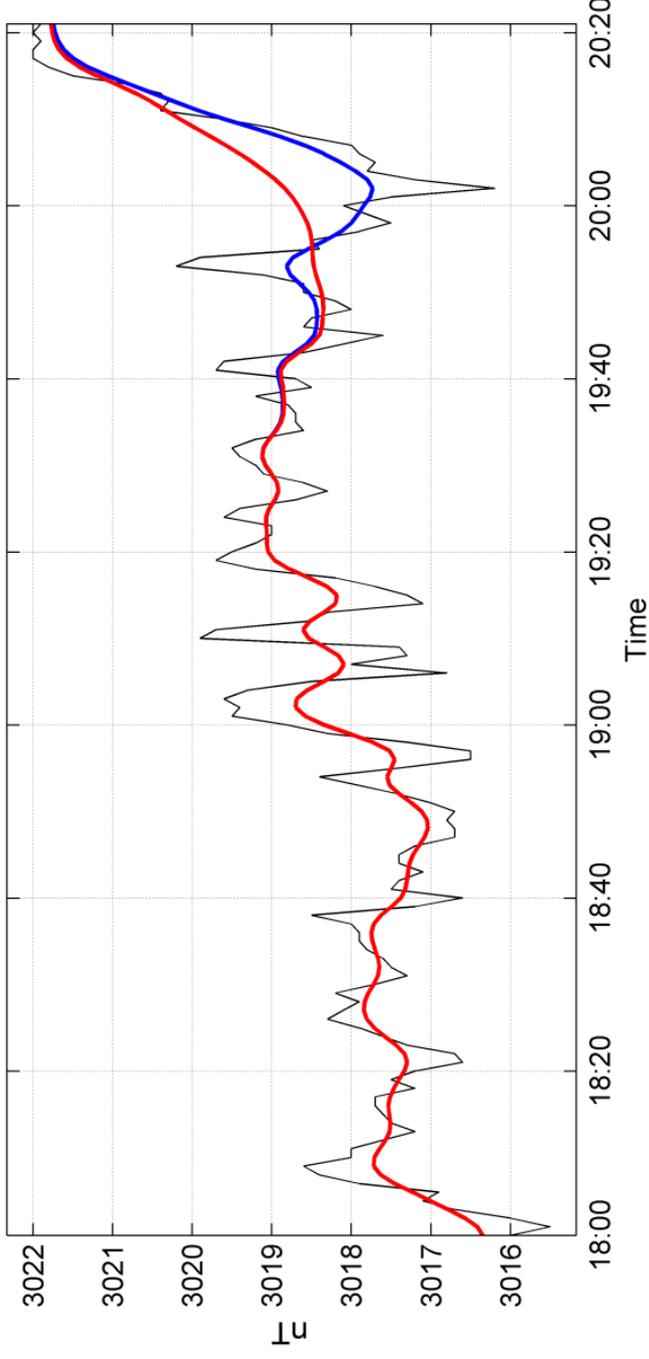


Figure 8. The points' weights in the period 19:50–20:10 are equal to 0.1.

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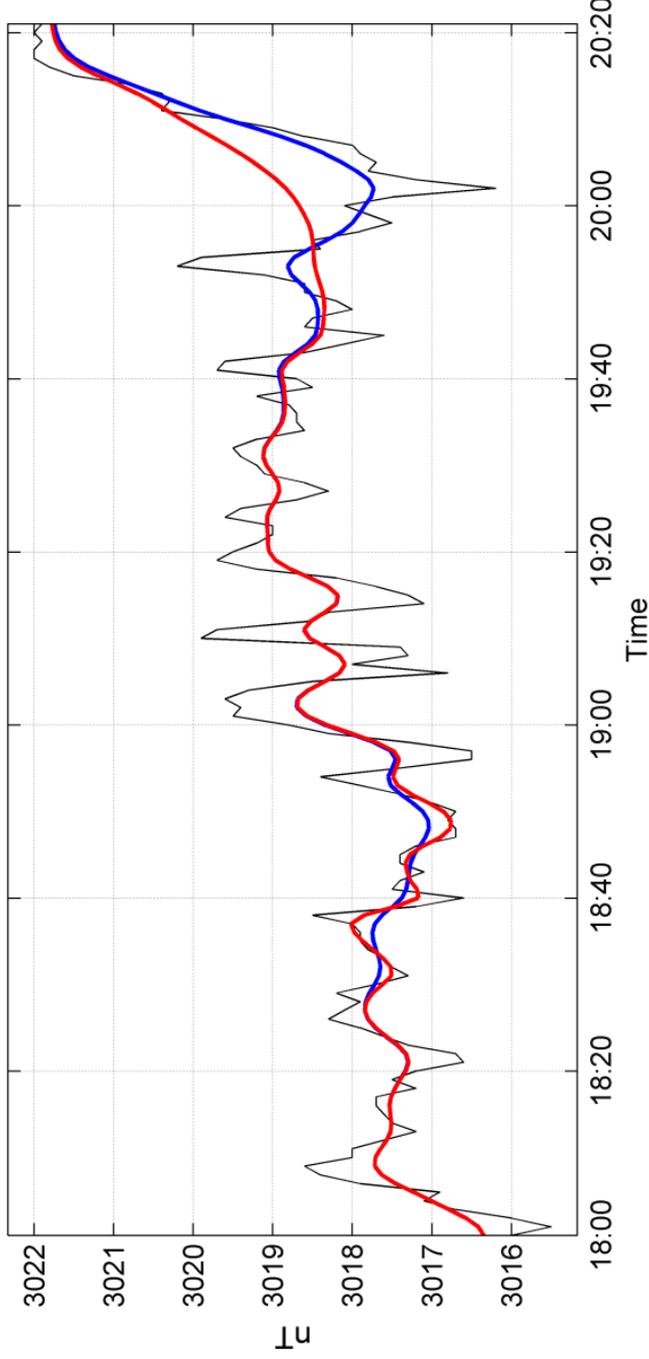


Figure 9. The points' weights in the period 18:30–18:50 are equal to 10.0, in the period 19:50–20:10 are equal to 0.1.

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