

“Space Data Processing: Making Sense of Experimental Data”

Topic 4

**"Process reconstruction free from
any constraints and assumptions"**

Tatiana Podladchikova Rupert Gerzer

Term 4, March 28 – May 27, 2016

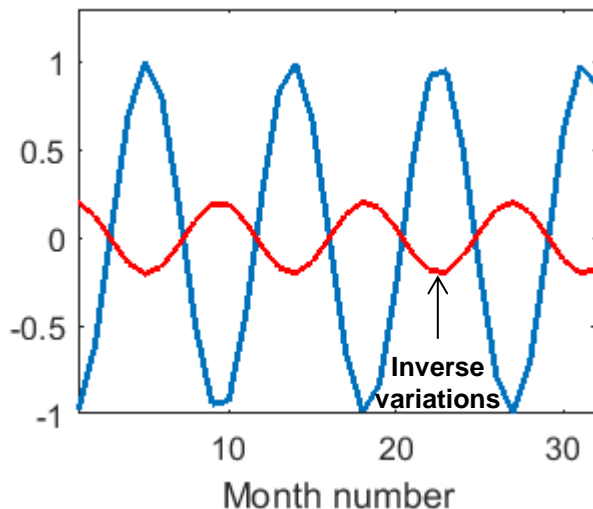
t.podladchikova@skoltech.ru

Analysis of running mean errors

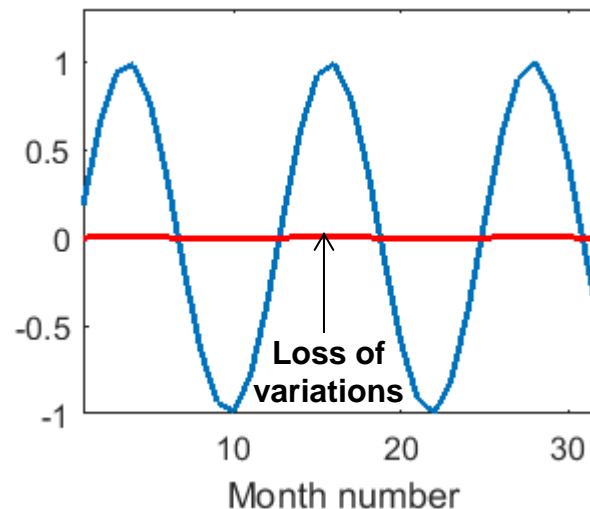
Running mean may significantly distort the dynamics of the process

— Measurements
— Smoothed, window size =13

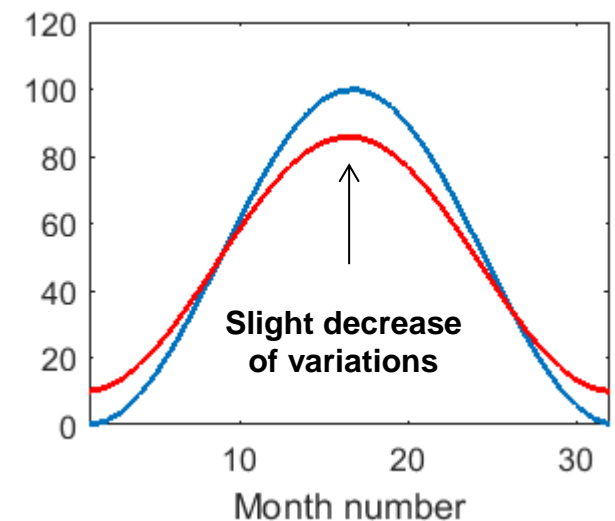
9-months variation



13-months variation



32-months variation

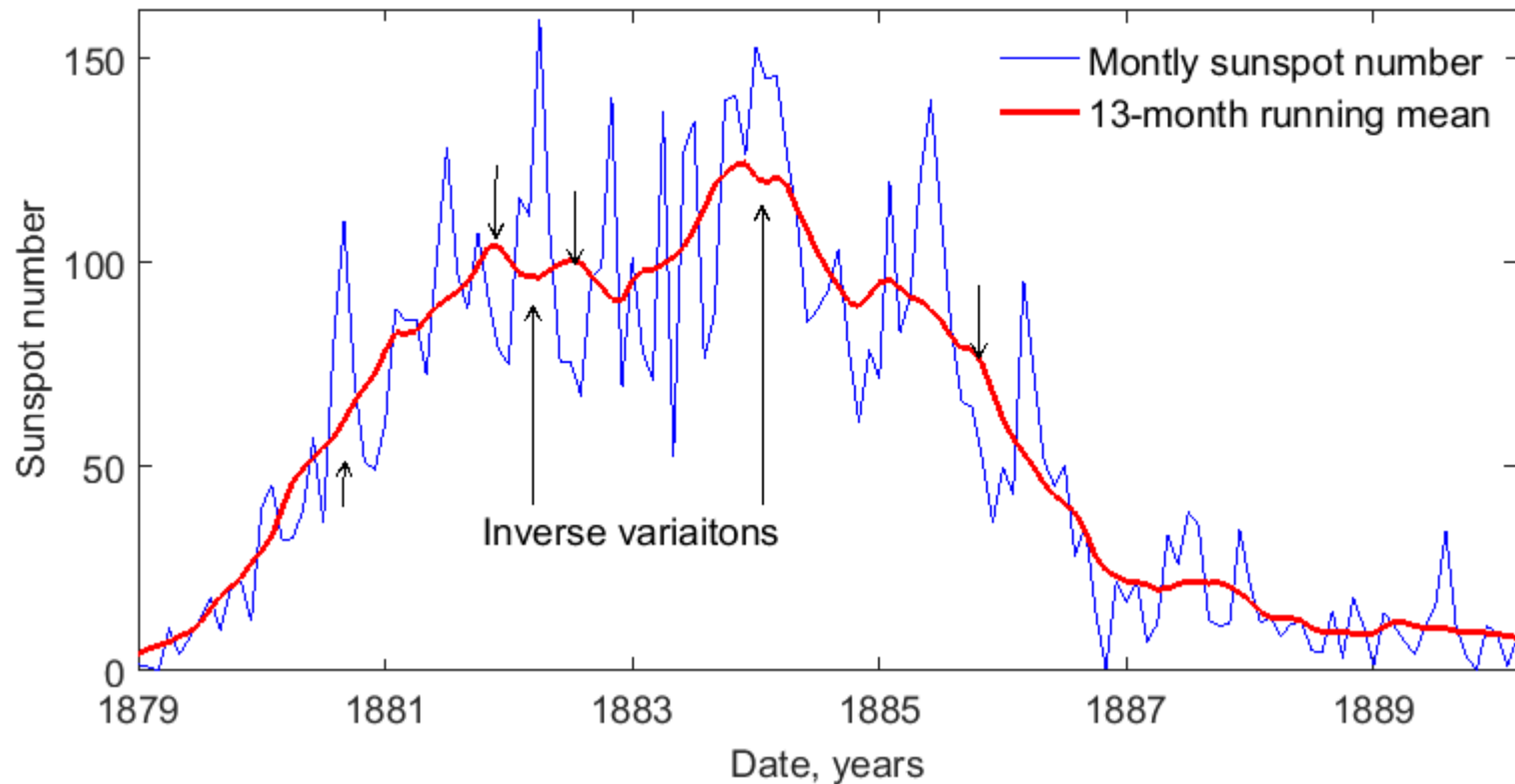


Inverse variations with periods from 6 to 12 months. Convex curve is replaced by concave curve and vice versa

Total loss of 6- and 12-month variations decreasing them to zero

Period greater than running window size (13 months). The process in general is not distorted

Distortion of physics in sunspot cycle 12



Performed analysis allows us to anticipate the errors of smoothing and getting false conclusions

Goals

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graph TD; Goals[Goals] --> Goal1[1 Reconstruct process free from any assumptions]; Goals --> Goal2[2 13-month optimal running mean];
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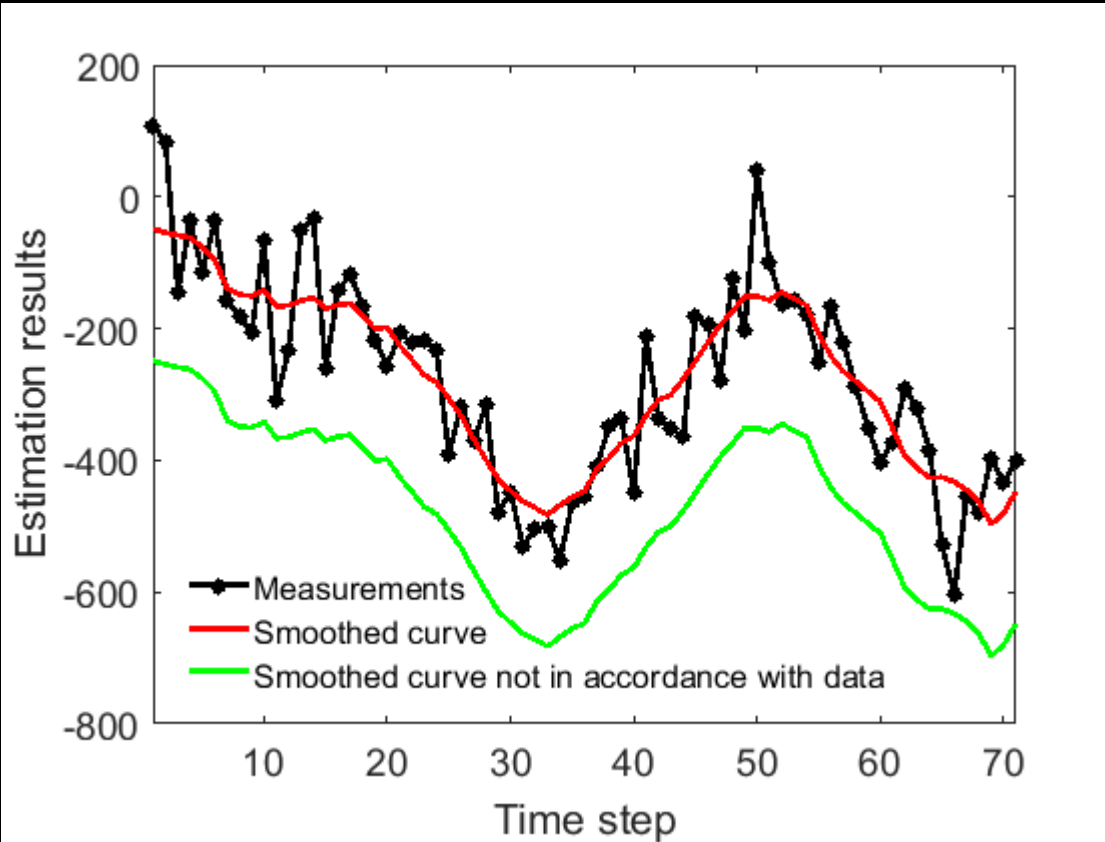
1

**Reconstruct
process free from
any assumptions**

2

**13-month optimal
running mean**

How reconstruct a process free from any assumptions?



Approximating curve \hat{X}_i has to be close to measurements z_i

1

Deviation indicator

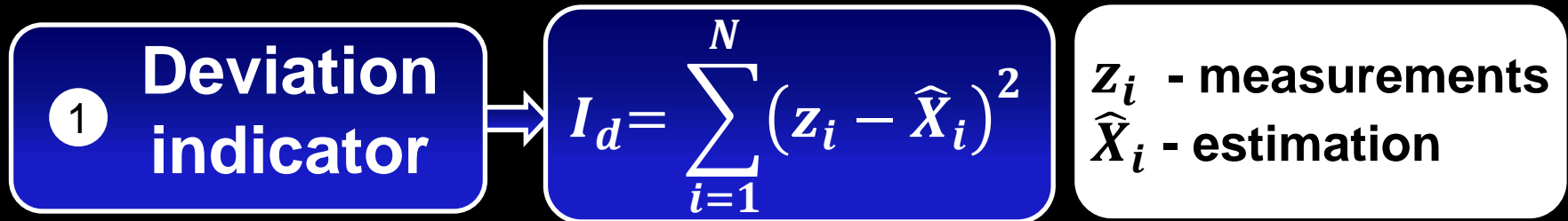


$$I_d = \sum_{i=1}^N (z_i - \hat{X}_i)^2$$

z_i - measurements
 \hat{X}_i - estimation

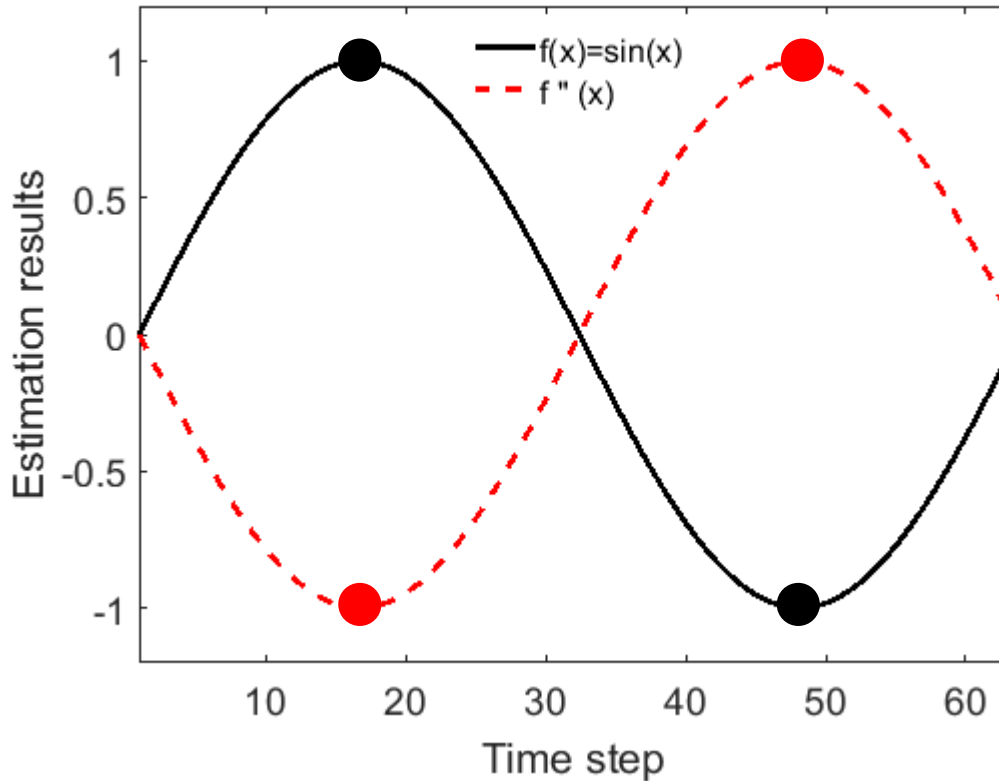
Goal 1

How reconstruct a process free from any assumptions?



Not enough to use only deviation indicator.
Additional criterion is needed

How reconstruct a process free from any assumptions?



Absolute value of second derivative is maximal at points of the greatest “variability” of approximating curve

Maximal rate of change of the process

2 Variability indicator

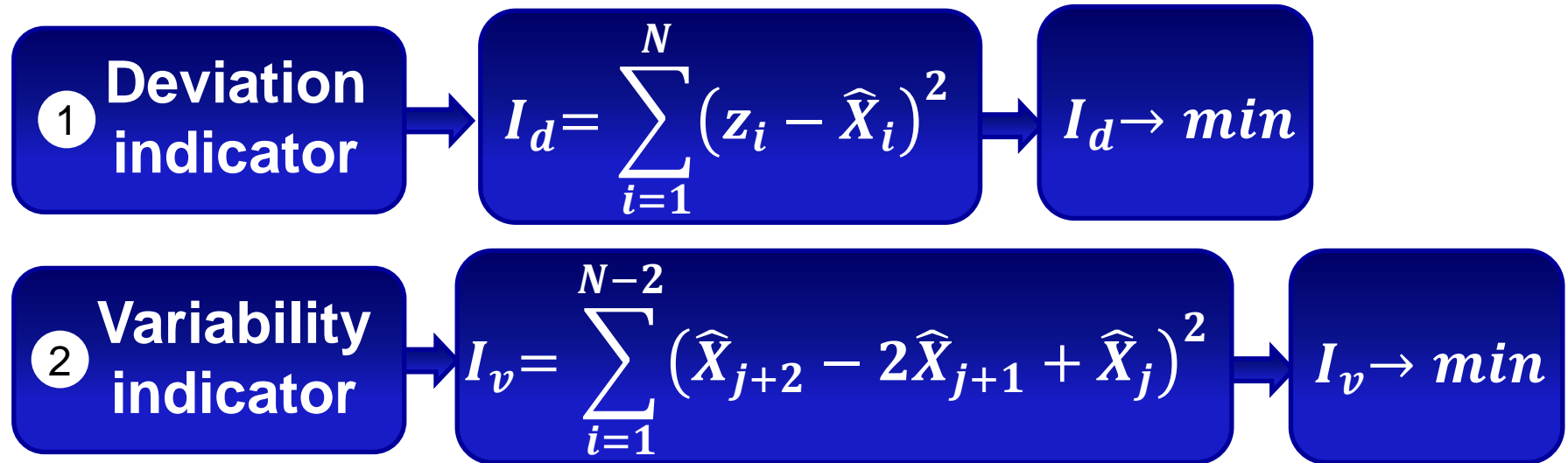
$$I_v = \sum_{i=1}^{N-2} (\hat{X}_{j+2} - 2\hat{X}_{j+1} + \hat{X}_j)^2$$

\hat{X}_i - estimation

Goal 1

Optimality criteria to find best approximation

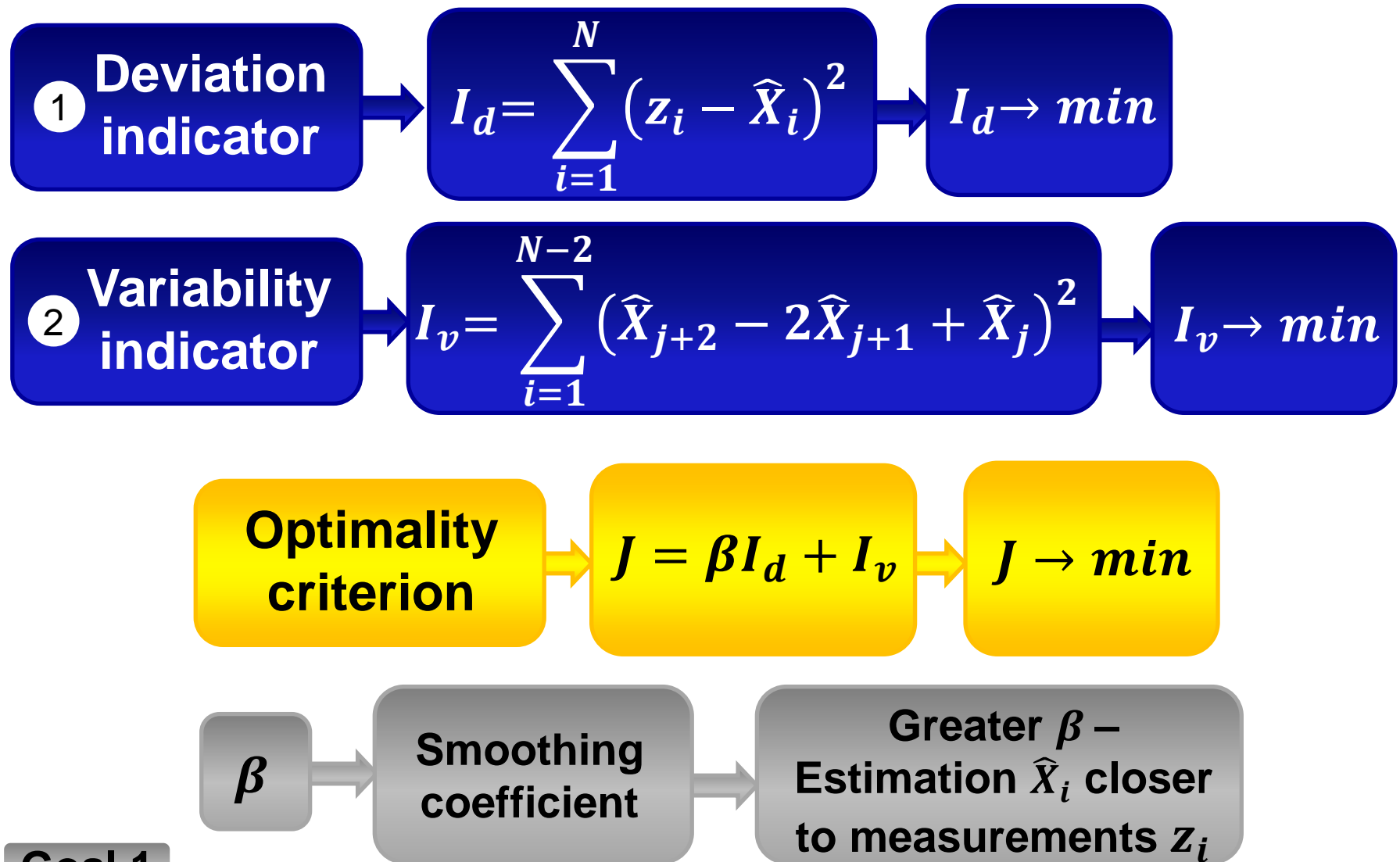
z_i - measurements, \hat{X}_i - estimation



Optimality
criterion
combining
 I_d and I_v ?

Optimality criteria to find best approximation

z_i - measurements, \hat{X}_i - estimation



Optimality criteria to find best approximation

$$J = \beta \sum_{i=1}^N (z_i - \hat{X}_i)^2 + \sum_{i=1}^{N-2} (\hat{X}_{j+2} - 2\hat{X}_{j+1} + \hat{X}_j)^2$$

Deviation
indicator I_d

Variability
indicator I_v

$J \rightarrow \min$

Fidelity to
measurements
 z_i

Smoothness
of \hat{X}_i

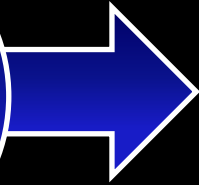
Balance

G. Bohlmann, 1899
E.T. Whittaker, 1923

Goal 1

Minimization of functional J to find best estimation

①
Let's
minimize
functional
 J



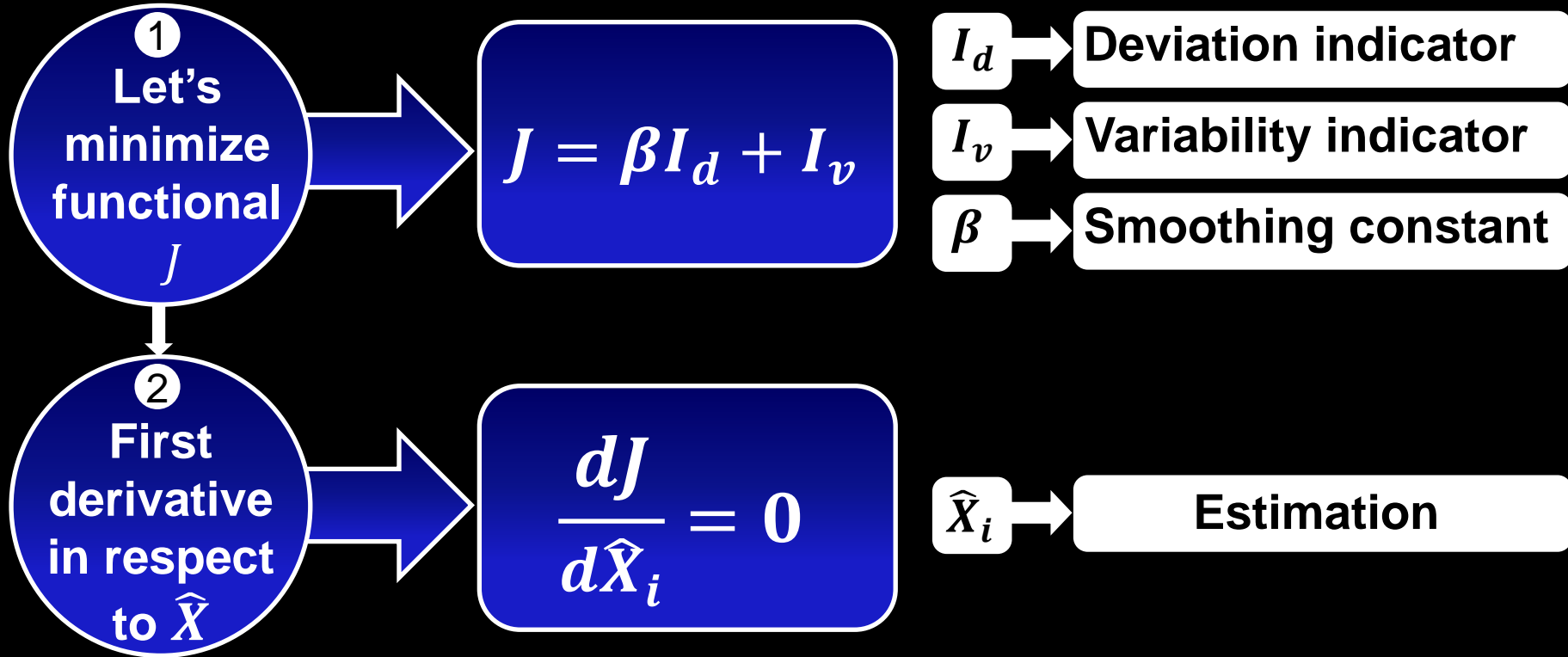
$$J = \beta I_d + I_v$$

I_d → Deviation indicator

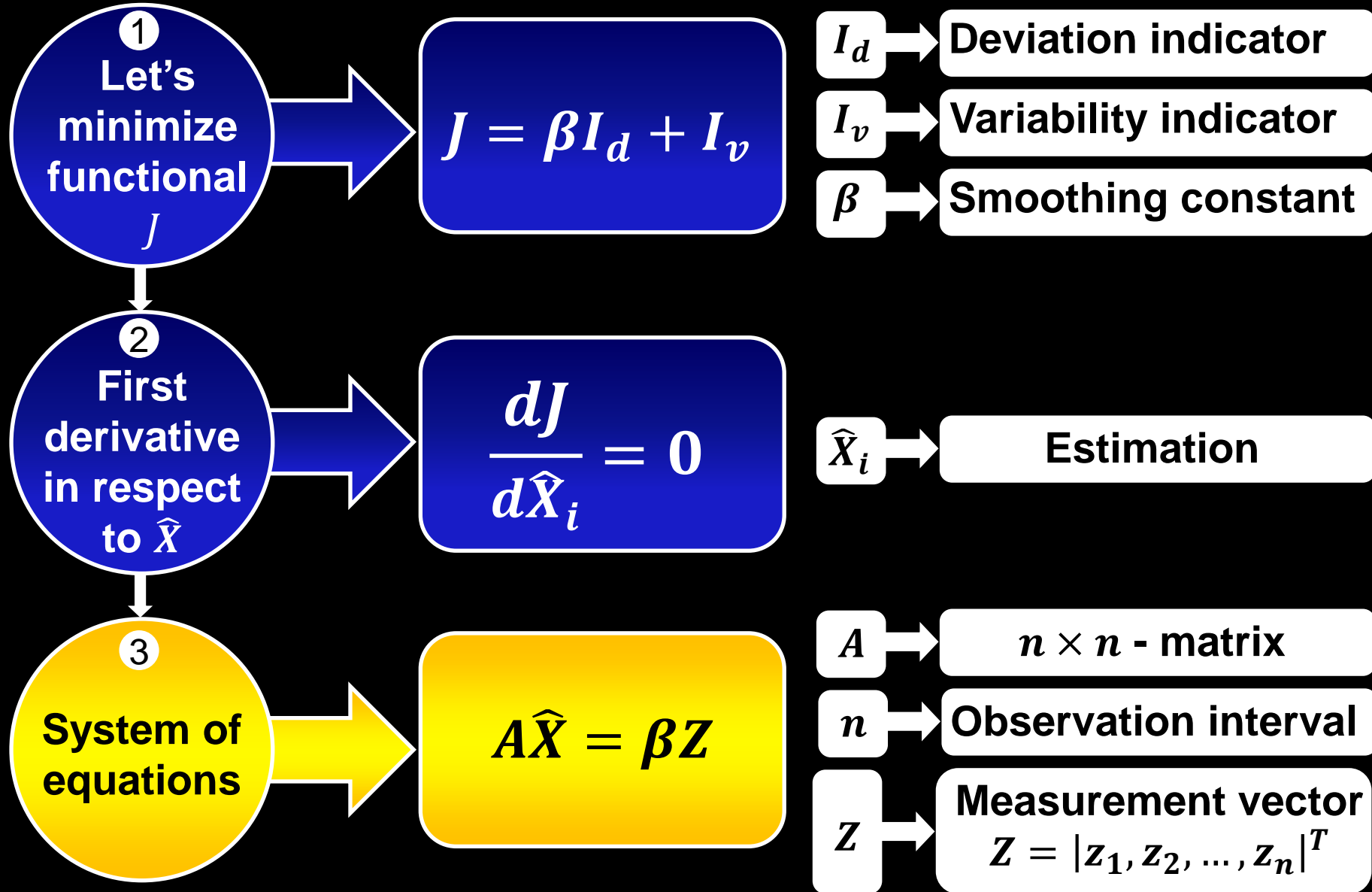
I_v → Variability indicator

β → Smoothing constant

Minimization of functional J to find best estimation

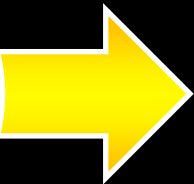


Minimization of functional J to find best estimation



Minimization of functional J to find best estimation

System of equations



$$A\hat{X} = \beta Z$$

n equations
with n unknowns

\hat{X}_i

Estimation

n

Observation interval

Z

Measurement vector
 $Z = [z_1, z_2, \dots, z_n]^T$

β

Smoothing constant

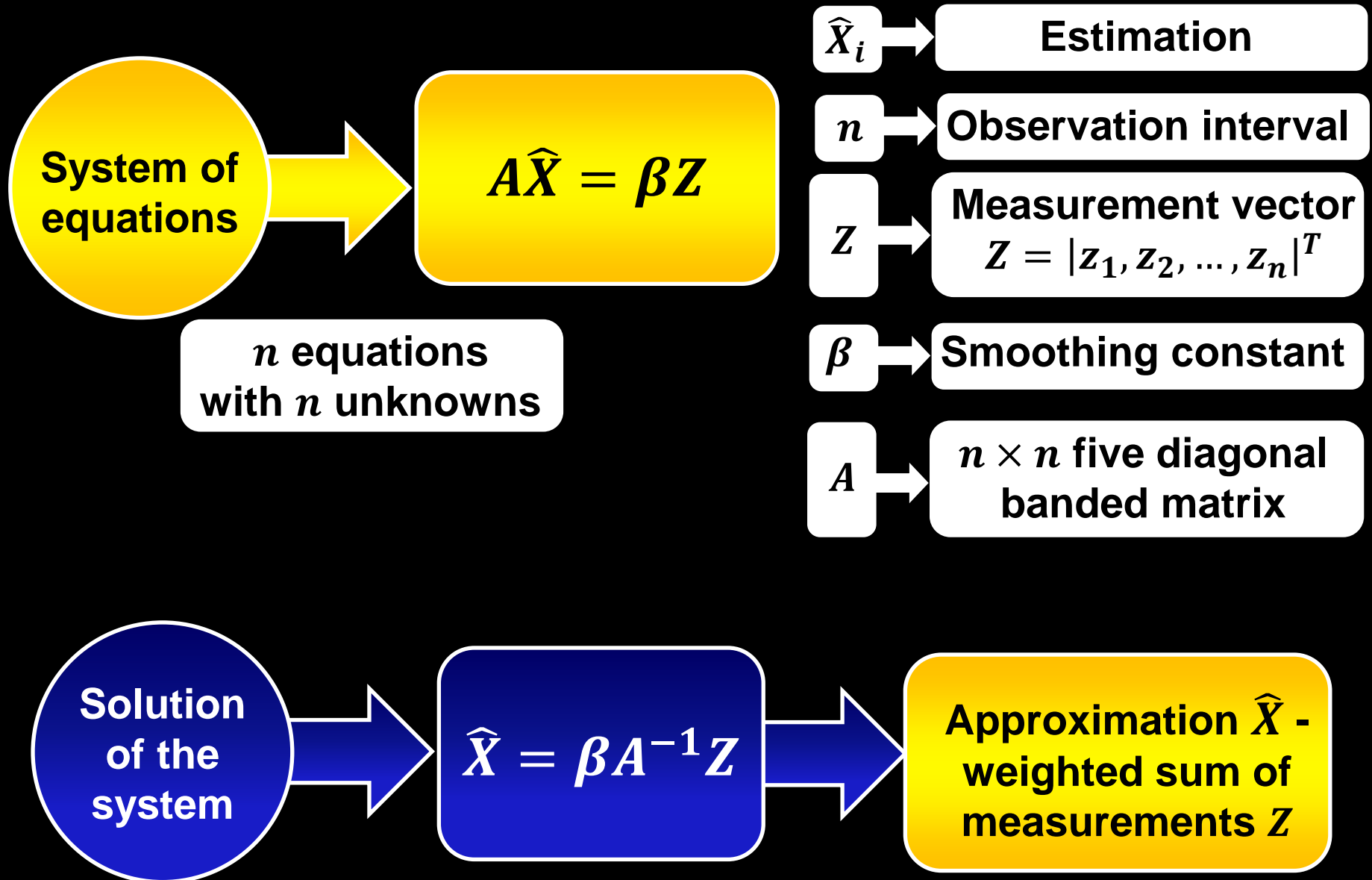
$$A = \begin{bmatrix} 1 + \beta & -2 & 1 & 0 & 0 & \dots & 0 \\ -2 & 5 + \beta & -4 & 1 & 0 & \dots & 0 \\ 1 & -4 & 6 + \beta & -4 & 1 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & \dots & 1 & -4 & 6 + \beta & -4 & 1 \\ 0 & \dots & 0 & 1 & -4 & 5 + \beta & -2 \\ 0 & \dots & 0 & 0 & 1 & -2 & 1 + \beta \end{bmatrix}$$

A

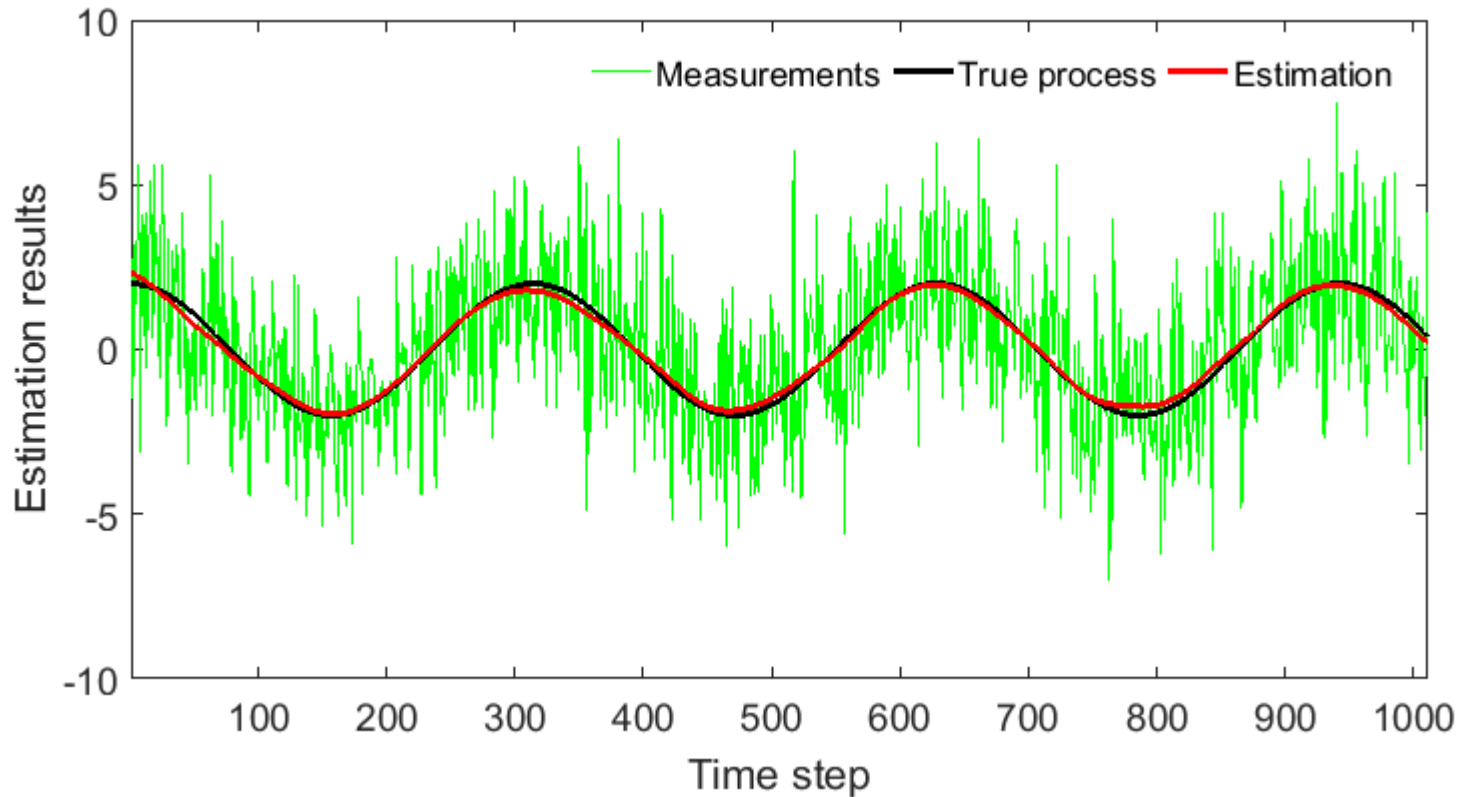
$n \times n$ five diagonal banded matrix

Goal 1

Minimization of functional J to find best estimation

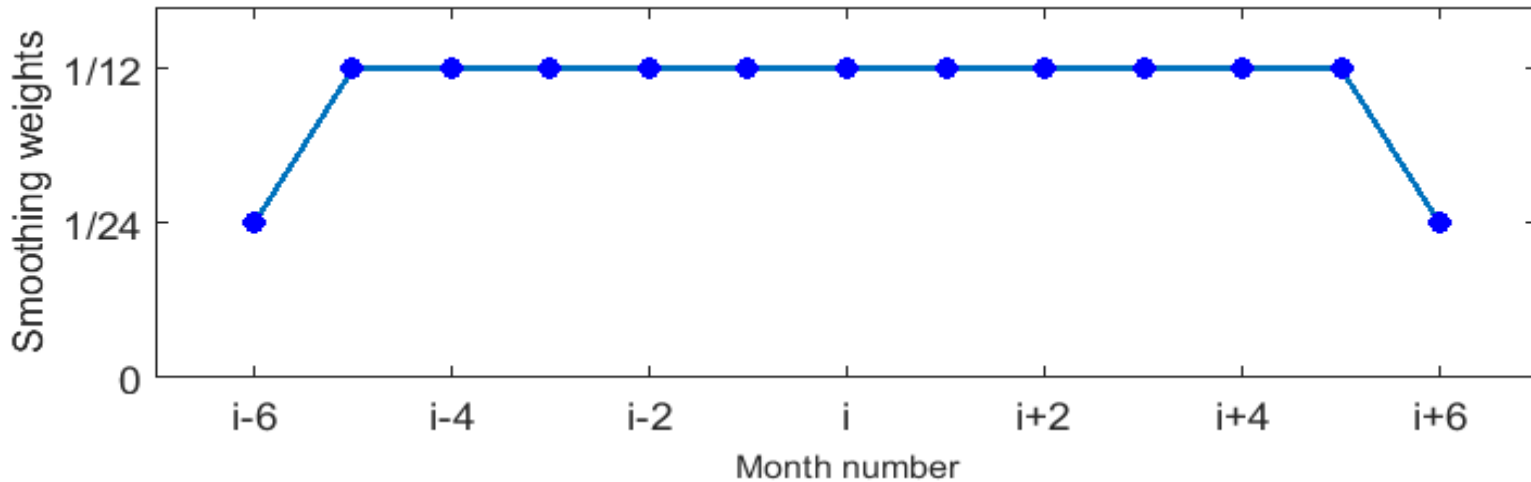


Reconstruction of a process free from any assumptions



13-month running mean

13-month sequent measurements Z



13-month running mean \hat{X}

$$\hat{X}_i = \frac{1}{24} z_{i-6} + \frac{1}{12} (z_{i-5} + z_{i-4} + \cdots + z_{i-1} + z_i + z_{i+1} + \cdots + z_{i+5}) + \frac{1}{24} z_{i+6}$$

13-month optimal running mean

13-month sequent
Measurements Z



$$Z = |z_{i-6}, z_{i-5}, \dots, z_i, \dots, z_{i+5}, z_{i+6}|^T$$

13-month optimal running mean

13-month sequent
Measurements Z



$$Z = |z_{i-6}, z_{i-5}, \dots, z_i, \dots, z_{i+5}, z_{i+6}|^T$$

13-month optimal
running mean \hat{X}



$$\hat{X} = |x_{i-6}, x_{i-5}, \dots, x_i, \dots, x_{i+5}, x_{i+6}|^T$$



?

13-month optimal running mean

13-month sequent
Measurements Z



$$Z = |z_{i-6}, z_{i-5}, \dots, z_i, \dots, z_{i+5}, z_{i+6}|^T$$

13-month optimal
running mean \hat{X}



$$\hat{X} = |x_{i-6}, x_{i-5}, \dots, x_i, \dots, x_{i+5}, x_{i+6}|^T$$



?

Optimal approximation \hat{X} -
weighted sum of measurements Z



$$\hat{X} = \beta A^{-1} Z$$

13-month optimal running mean

Optimal approximation \hat{X} -
weighted sum of measurements Z

$$\hat{X} = \beta A^{-1} Z$$

$1 + \beta$	-2	1	0	0	0	0	0	0	0	0	0	0	0
-2	$5 + \beta$	-4	1	0	0	0	0	0	0	0	0	0	0
1	-4	$6 + \beta$	-4	1	0	0	0	0	0	0	0	0	0
0	1	-4	$6 + \beta$	-4	1	0	0	0	0	0	0	0	0
0	0	1	-4	$6 + \beta$	-4	1	0	0	0	0	0	0	0
0	0	0	1	-4	$6 + \beta$	-4	1	0	0	0	0	0	0
0	0	0	0	1	-4	$6 + \beta$	-4	1	0	0	0	0	0
0	0	0	0	0	1	-4	$6 + \beta$	-4	1	0	0	0	0
0	0	0	0	0	0	1	-4	$6 + \beta$	-4	1	0	0	0
0	0	0	0	0	0	0	1	-4	$6 + \beta$	-4	1	0	0
0	0	0	0	0	0	0	0	1	-4	$6 + \beta$	-4	1	0
0	0	0	0	0	0	0	0	0	1	-4	$6 + \beta$	-4	1
0	0	0	0	0	0	0	0	0	0	1	-4	$5 + \beta$	-2
0	0	0	0	0	0	0	0	0	0	0	1	-2	$1 + \beta$

A

→ 13×13 five diagonal banded matrix

13-month optimal running mean

Optimal approximation \hat{X} -
weighted sum of measurements Z

$$\hat{X} = \beta A^{-1} Z$$

$$\hat{X} = \begin{bmatrix} x_{i-6} \\ x_{i-5} \\ \dots \\ x_i \\ \dots \\ x_{i+5} \\ x_{i+6} \end{bmatrix} = \beta A^{-1} \begin{bmatrix} z_{i-6} \\ z_{i-5} \\ \dots \\ z_i \\ \dots \\ z_{i+5} \\ z_{i+6} \end{bmatrix}$$

$A \rightarrow 13 \times 13$ five diagonal banded matrix

$c_{ij} \rightarrow$ Elements of matrix A^{-1}

13-month optimal running mean

Optimal approximation \hat{X} -
weighted sum of measurements Z

$$\hat{X} = \beta A^{-1} Z$$

$$\hat{X} = \begin{bmatrix} x_{i-6} \\ x_{i-5} \\ \dots \\ x_i \\ \dots \\ x_{i+5} \\ x_{i+6} \end{bmatrix} = \beta \begin{bmatrix} c_{1,1} & c_{1,2} & \dots & c_{1,7} & \dots & c_{1,12} & c_{1,13} \\ c_{2,1} & c_{2,2} & \dots & c_{2,7} & \dots & c_{2,12} & c_{2,13} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ c_{7,1} & c_{7,2} & \dots & c_{7,7} & \dots & c_{7,12} & c_{7,13} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ c_{12,1} & c_{12,2} & \dots & c_{12,7} & \dots & c_{12,12} & c_{12,13} \\ c_{13,1} & c_{13,2} & \dots & c_{13,7} & \dots & c_{13,12} & c_{13,13} \end{bmatrix} \begin{bmatrix} z_{i-6} \\ z_{i-5} \\ \dots \\ z_i \\ \dots \\ z_{i+5} \\ z_{i+6} \end{bmatrix}$$

$A \rightarrow 13 \times 13$ five diagonal banded matrix

$c_{ij} \rightarrow$ Elements of matrix A^{-1}

13-month optimal running mean

Optimal approximation \hat{X} -
weighted sum of measurements Z

$$\hat{X} = \beta A^{-1} Z$$

$$\hat{X} = \begin{bmatrix} x_{i-6} \\ x_{i-5} \\ \dots \\ x_i \\ \dots \\ x_{i+5} \\ x_{i+6} \end{bmatrix} = \beta \begin{bmatrix} c_{1,1} & c_{1,2} & \dots & c_{1,7} & \dots & c_{1,12} & c_{1,13} \\ c_{2,1} & c_{2,2} & \dots & c_{2,7} & \dots & c_{2,12} & c_{2,13} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ c_{7,1} & c_{7,2} & \dots & c_{7,7} & \dots & c_{7,12} & c_{7,13} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ c_{12,1} & c_{12,2} & \dots & c_{12,7} & \dots & c_{12,12} & c_{12,13} \\ c_{13,1} & c_{13,2} & \dots & c_{13,7} & \dots & c_{13,12} & c_{13,13} \end{bmatrix} \begin{bmatrix} z_{i-6} \\ z_{i-5} \\ \dots \\ z_i \\ \dots \\ z_{i+5} \\ z_{i+6} \end{bmatrix}$$

$A \rightarrow 13 \times 13$ five diagonal banded matrix

$c_{ij} \rightarrow$ Elements of matrix A^{-1}

How to get estimation x_i at time i ?

13-month optimal running mean

Optimal approximation \hat{X} -
weighted sum of measurements Z

$$\hat{X} = \beta A^{-1} Z$$

$$\hat{X} = \begin{bmatrix} x_{i-6} \\ x_{i-5} \\ \dots \\ x_i \\ \dots \\ x_{i+5} \\ x_{i+6} \end{bmatrix} = \beta \begin{bmatrix} c_{1,1} & c_{1,2} & \dots & c_{1,7} & \dots & c_{1,12} & c_{1,13} \\ c_{2,1} & c_{2,2} & \dots & c_{2,7} & \dots & c_{2,12} & c_{2,13} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ c_{7,1} & c_{7,2} & \dots & c_{7,7} & \dots & c_{7,12} & c_{7,13} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ c_{12,1} & c_{12,2} & \dots & c_{12,7} & \dots & c_{12,12} & c_{12,13} \\ c_{13,1} & c_{13,2} & \dots & c_{13,7} & \dots & c_{13,12} & c_{13,13} \end{bmatrix} \begin{bmatrix} z_{i-6} \\ z_{i-5} \\ \dots \\ z_i \\ \dots \\ z_{i+5} \\ z_{i+6} \end{bmatrix}$$

$A \rightarrow 13 \times 13$ five diagonal banded matrix

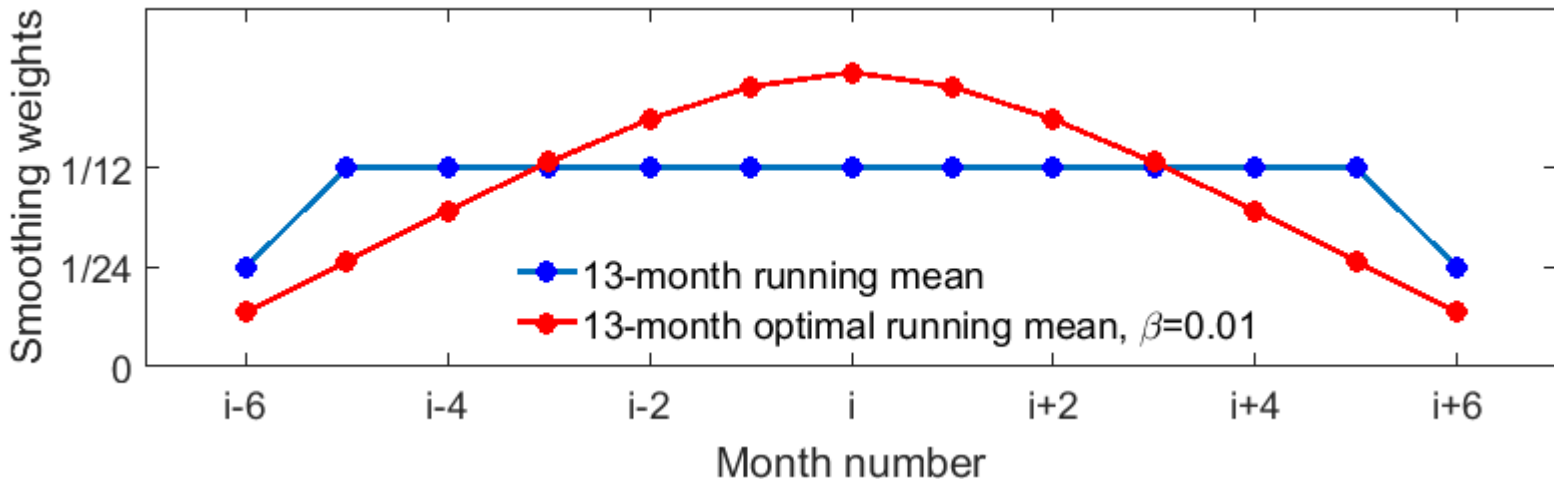
$c_{ij} \rightarrow$ Elements of matrix A^{-1}

$$x_i = \beta C(\beta) Z$$

$$C(\beta) = [c_{7,1}, c_{7,2}, \dots, c_{7,7}, \dots, c_{7,12}, c_{7,13}]$$

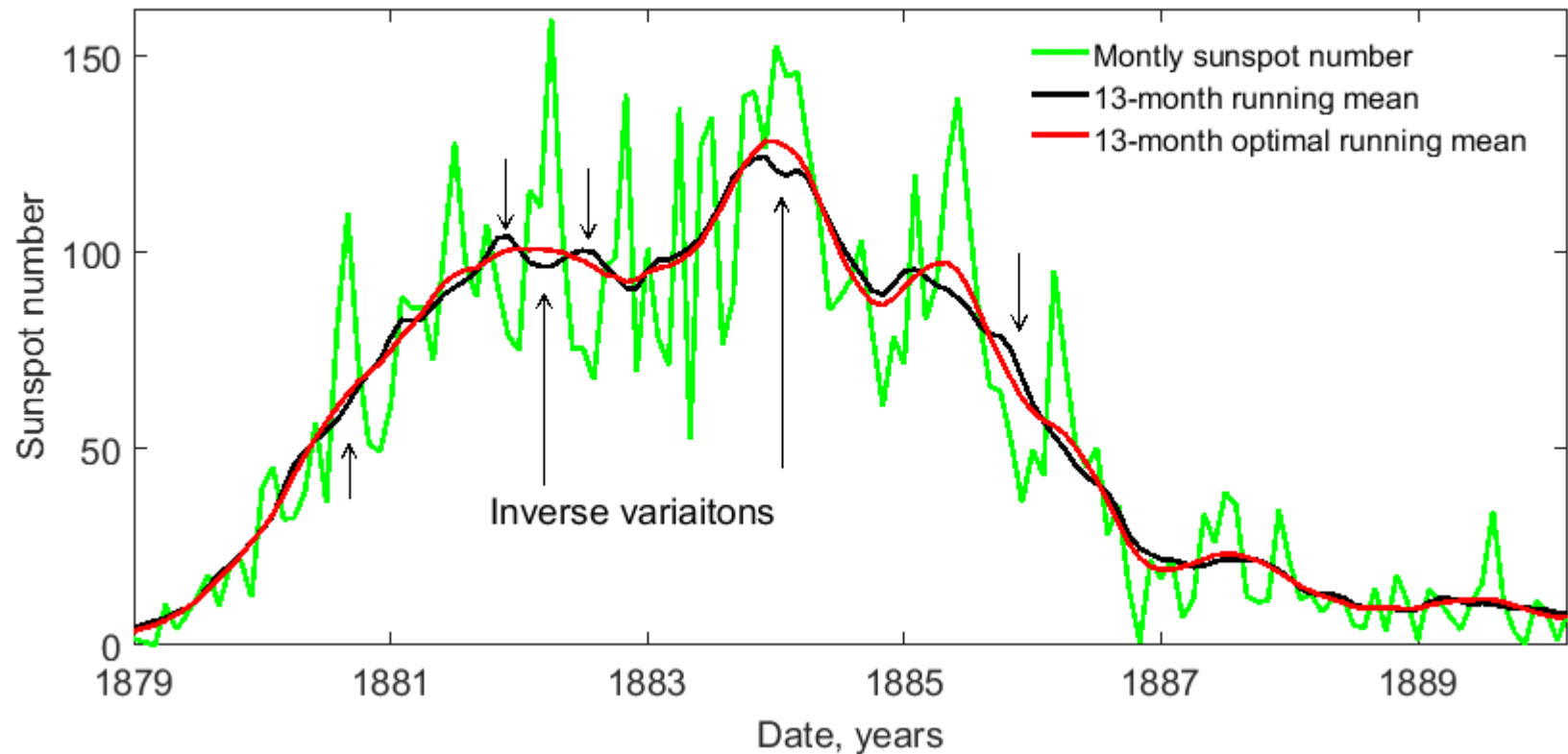
Smoothing weights in 13-month running mean and 13-month optimal running mean

13-month sequent measurements Z



The optimal weights monotonously increase when measurements approach step i

Approximation of sunspot cycle 12



The optimal filter provides more adequate presentation of the sunspot cycle and doesn't distort the short-term variations of sunspot numbers

Comparative analysis of optimal and non-optimal techniques

To analyze qualitatively advantages of optimal smoothing technique compared to the 13-month running mean



Determine and compare deviation indicator I_d and variability indicator I_v