

**L. Yaroslavsky. Lectures on Selected Topics Pt. 5. Local adaptive filters for image restoration and enhancement**

**Lect. 12. Local adaptive nonlinear filters.**

Local criteria of image quality:  $AVLOSS(k,l) = AV \left\{ \sum_{m,n} LOC(m,n;a_{k,l}) LOSS(a_{m,n}, \hat{a}_{m,n}) \right\}$

Locality function  $LOC(\cdot)$  and loss function  $LOSS(\cdot)$  and a priori knowledge on images

Linear filters:  $\hat{a}_{k,l} = \lambda_0 a_{k,l}^{inp} + \bar{a}_{k,l}$ ;  $\bar{a}_{k,l}$  is a weighted average over a spatial neighborhood

Nonlinear filters (Fig. 1):  $\{a_{k,l}^{inp}\} \rightarrow \hat{a}_{k,l} : \hat{a}_{k,l} = ESTM(NBH(a_k))$

Rank filters:  $\hat{a}_{k,l} = F_{hst}(a_{k,l}^{inp})$  where  $F_{hst}(\cdot)$  is a non-linear function determined by the local histogram over a certain neighborhood of the pixel  $(k,l)$ .

Neighborhoods and pixel attributes:

In terms of pixel position: Spatial  $S$ -neighborhoods

In terms of pixel values ( $V$ -neighborhoods):

$$EV\{a_{k,l}\} = \{a_{m,n} : a_{k,l} - \varepsilon_v^- \leq a_{m,n} \leq a_{k,l} + \varepsilon_v^+\} \quad KNV(a_{k,l}) = \left\{ a_{m,n}^p : \sum_{p=1}^K |a_{k,l} - a_{m,n}^p| = \min_{m,n} \right\}$$

In terms of pixel rank (position in the variational row) ( $R$ -neighborhoods):

$$ER\{a_{k,l}\} = \{a_{m,n} : R(a_{k,l}) - \varepsilon_r^- \leq R(a_{m,n}) \leq R(a_{k,l}) + \varepsilon_r^+\}; \quad KNR(a_{k,l}) = \left\{ a_{m,n}^p : \sum_{p=1}^K |R(a_{k,l}) - R(a_{m,n}^p)| = \min_{m,n} \right\}$$

In terms of histogram features (H-neighborhoods): **Cluster**, ( $CL$ -neighborhood): subset of pixels that belong to the same cluster, or mode, of the histogram as that of the central pixel.

In terms of image function geometrical features.

Other possible neighborhoods

Neighborhood operations (estimators):

**MEAN(NBH)** - arithmetic mean over the neighborhood  $NBH$ ;

$$\text{MEAN}(NBH) = \arg \min \left( \sum_k |a_k - \hat{a}|^2 \right); \quad \text{MEDN}(NBH) = \arg \min \left( \sum_k |a_k - \hat{a}| \right)$$

**ROS(NBH)** -  $R$ -th Order Statistics over the neighborhood; special cases:

**MEDN(NBH)**- median; **MAX(NBH)** - maximum; **MIN(NBH)** - minimum;

**MODE(NBH)** - a value that corresponds to the highest maximum of the histogram over the neighborhood;

**RAND(NBH)** - pseudo-random number taken from the same distribution as the histogram over the neighborhood;

**RANK(NBH)** - position of the central pixel of the window in the variational row over the neighborhood;

**SIZE(NBH)** - size (in pixels or gray levels, whatever is appropriate) of the neighborhoods.

Iterative nature of local adaptive nonlinear filters:

$$\hat{a}_k^{(t)} = ESTM(NBH^{(t-1)})$$

**TYPICAL ALGORITHMS:**

Rank filters for smoothing additive and impulse noise and image segmentation :

$$\hat{a}_{k,l}^{(t)} = SMTH(NBH(\hat{a}_{k,l}^{(t-1)})) \quad \text{and} \quad \hat{a}_{k,l}^{(t)} = \begin{cases} \hat{a}_{k,l}^{(t-1)}, & \text{if } |\hat{a}_{k,l}^{(t-1)} - SMTH(NBH(\hat{a}_{k,l}^{(t-1)}))| \leq thr \\ SMTH(NBH(\hat{a}_{k,l}^{(t-1)})), & \text{otherwise} \end{cases}$$

where  $SMTH$  is a smoothing operations (**MEAN**, **MED**, **ROS**, **MODE** or **RAND**),  $thr$  is a detection threshold  $t$  is number of the iteration.

Rank order filters for local contrast enhancement:

Unsharp masking:  $\hat{a}_{k,l} = G(a_{k,l} - SMTH(NBH(a_{k,l})))$ , where  $G$  is an enhancement coefficient,

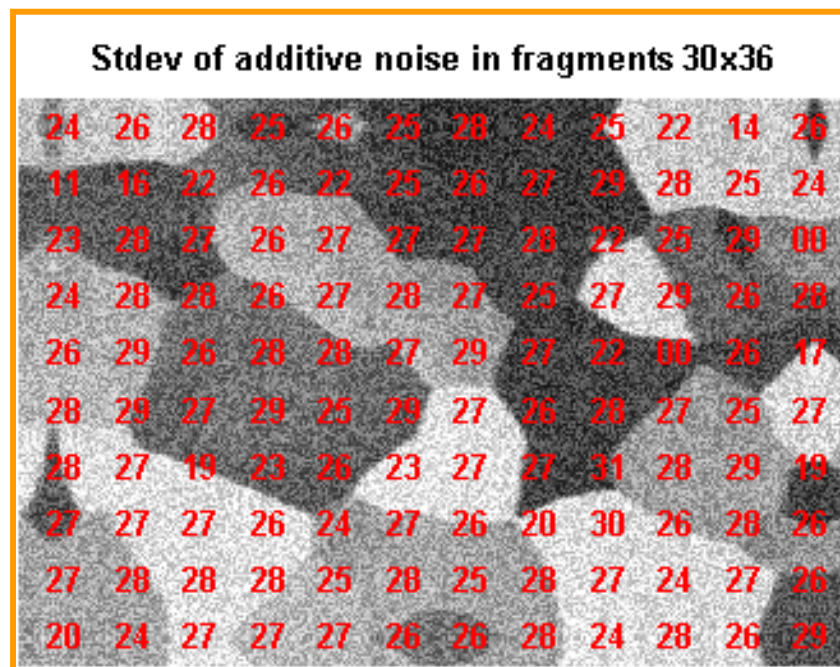
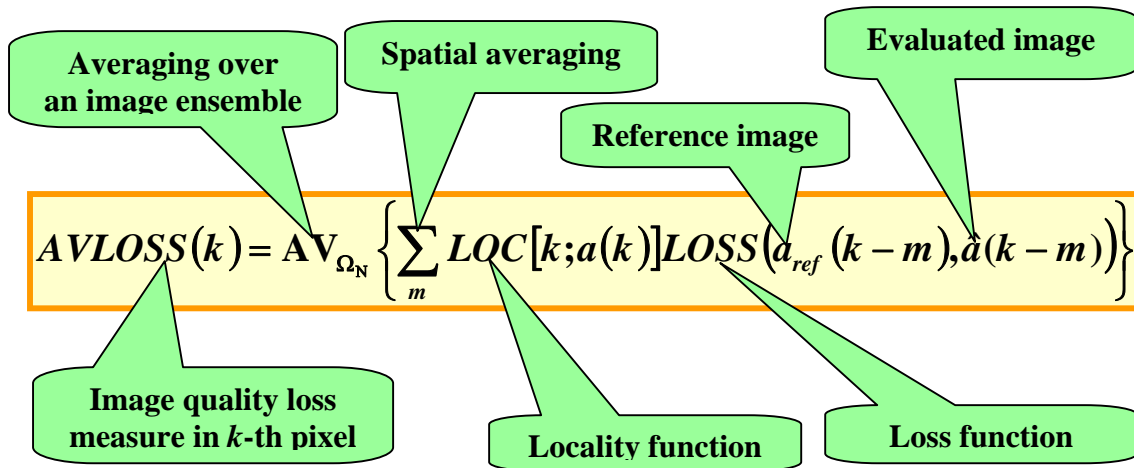
Local histogram equalization:  $\hat{a}_{k,l} = \text{RANK}(NBH(a_{k,l}))$ ; Local histogram  $p$ -equalization.

Possible generalizations and implementation issues:

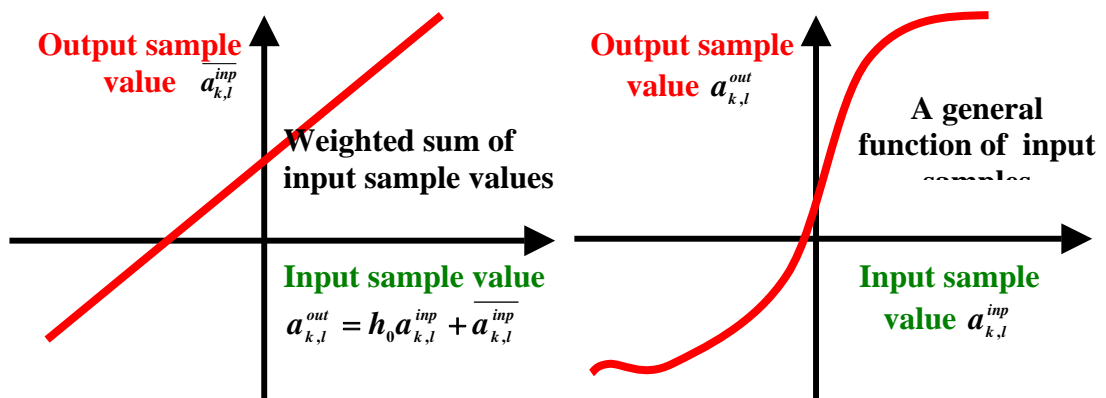
Weighted histograms; nonlinearly transformed histograms; new types of neighborhoods; extension of the set of basic operations.

Implementation of local adaptive filters in computers, optics and neuro-morphic parallel networks.

## Local criteria of image processing quality



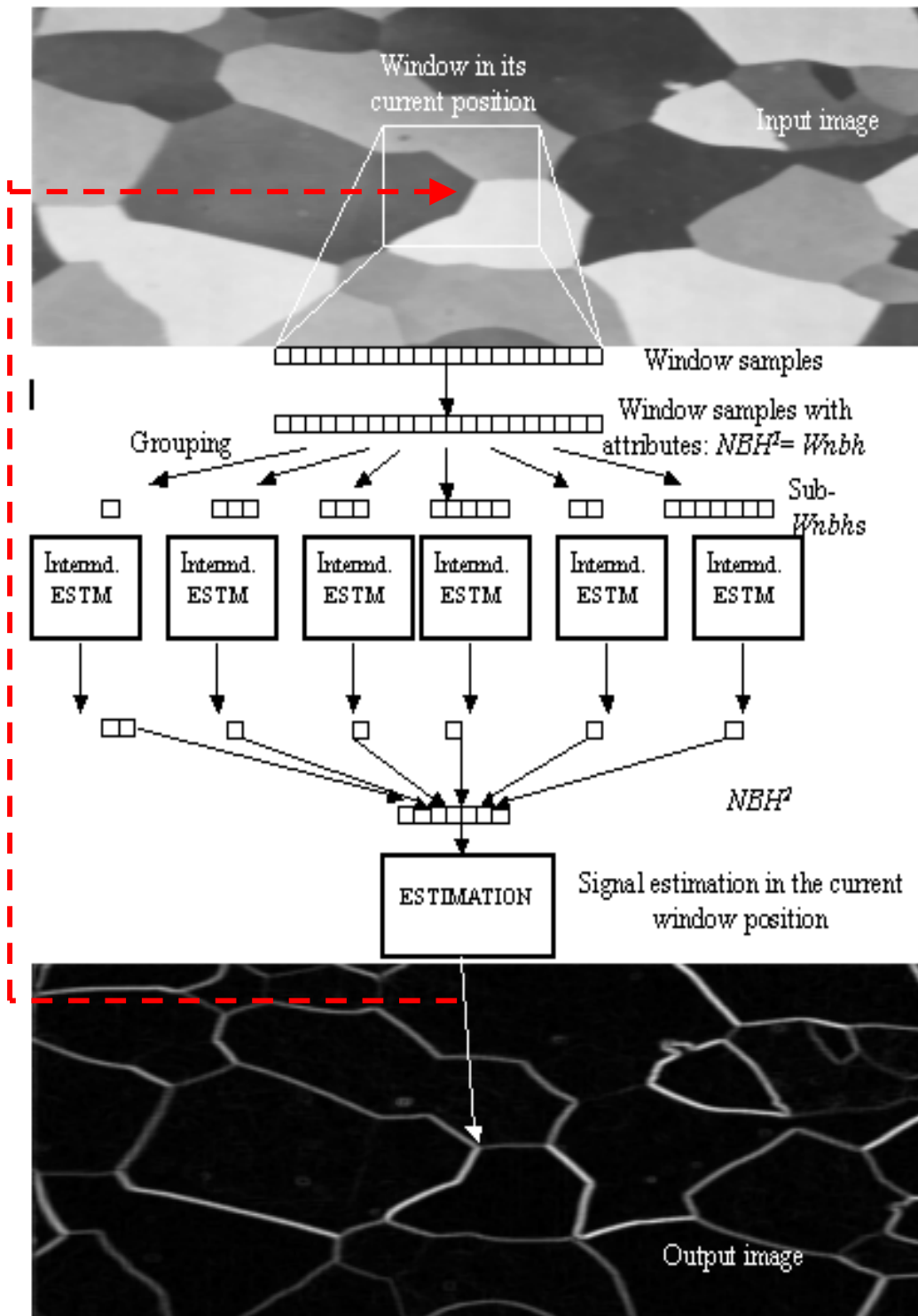
Noisevar\_disp.m



Linear and nonlinear filtering: the principle

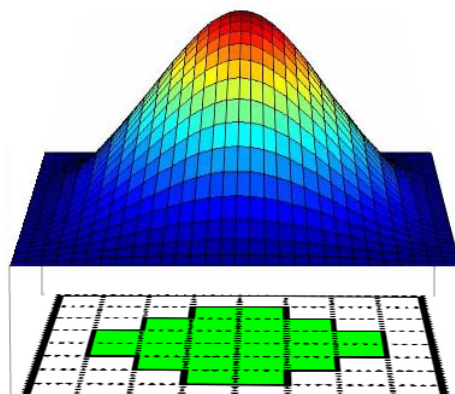
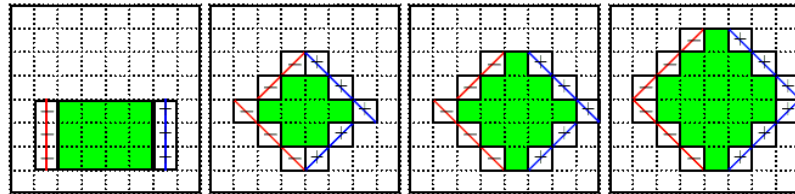


John Wilder Tukey, 1915-2000



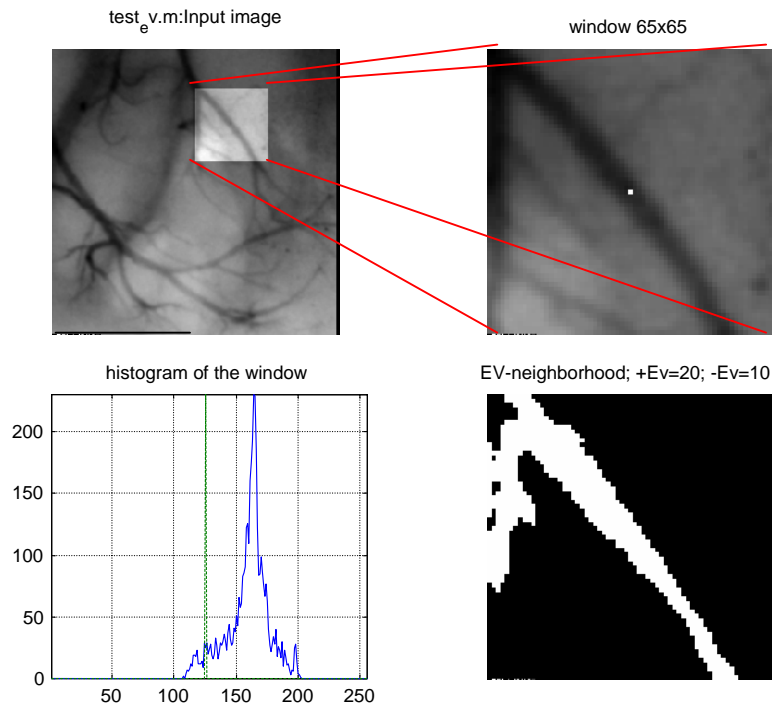
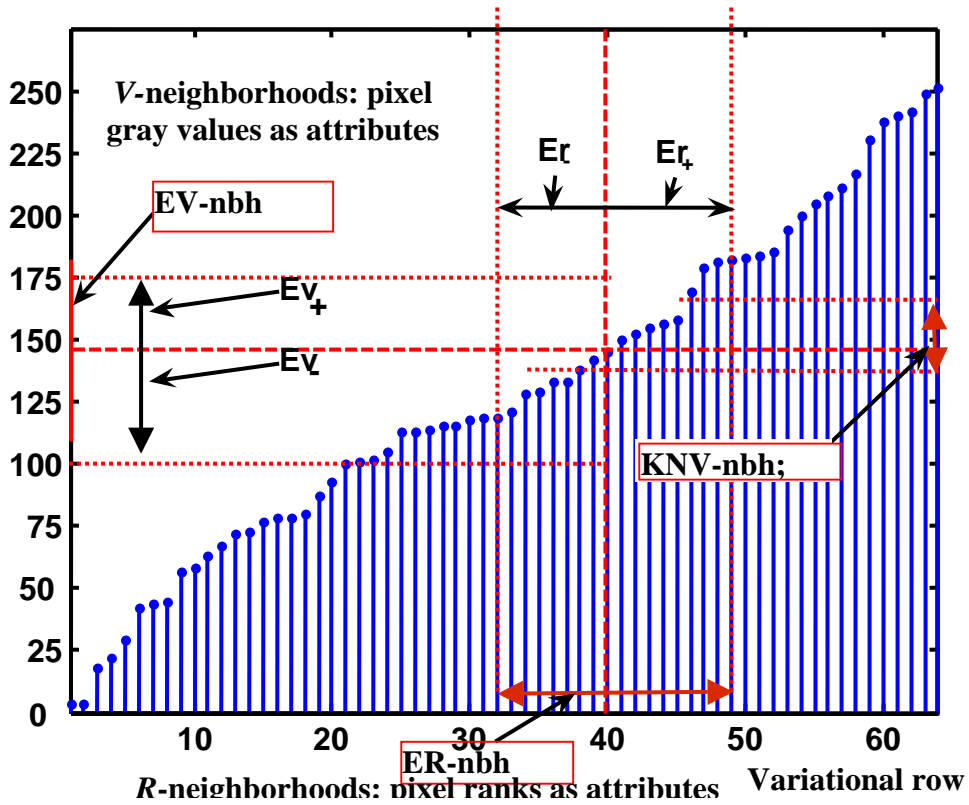
**A unified representation of multi stage nonlinear filters (on an example of  $NBH^2$ -filtering) and schematic diagram of non-linear filtering**

## S-neighborhoods



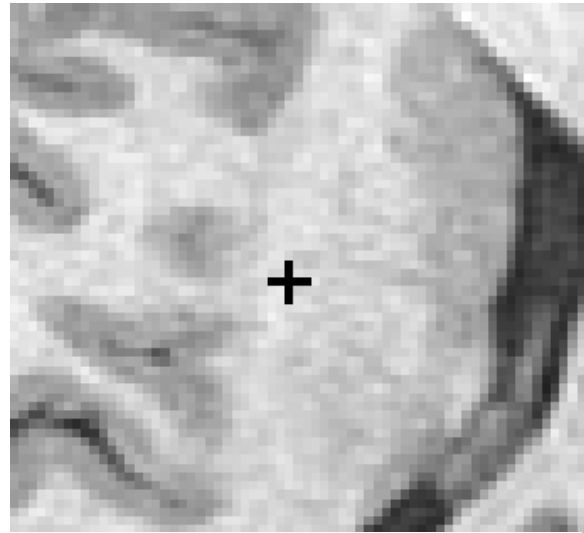
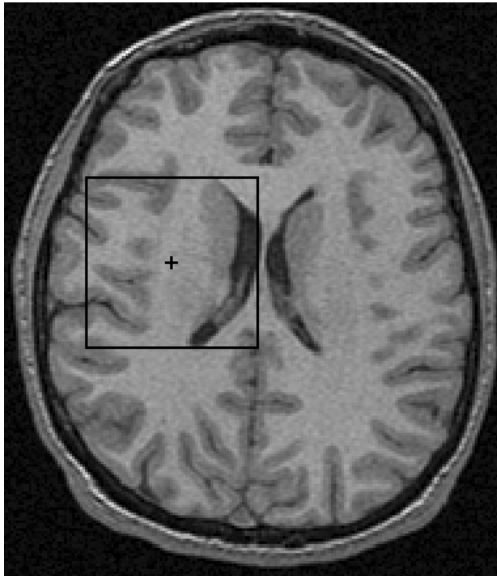
Weighted spatial neighborhood

### EV- and ER- neighborhoods

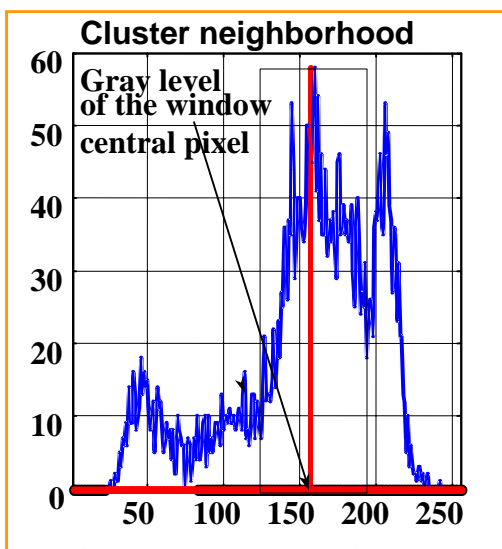


**EV-neighborhood: an illustrative example**

## H-neighborhoods: pixel cardinality as an attribute



Cluster-Nbh



Histogram over the window



## Information content in image global and local histograms

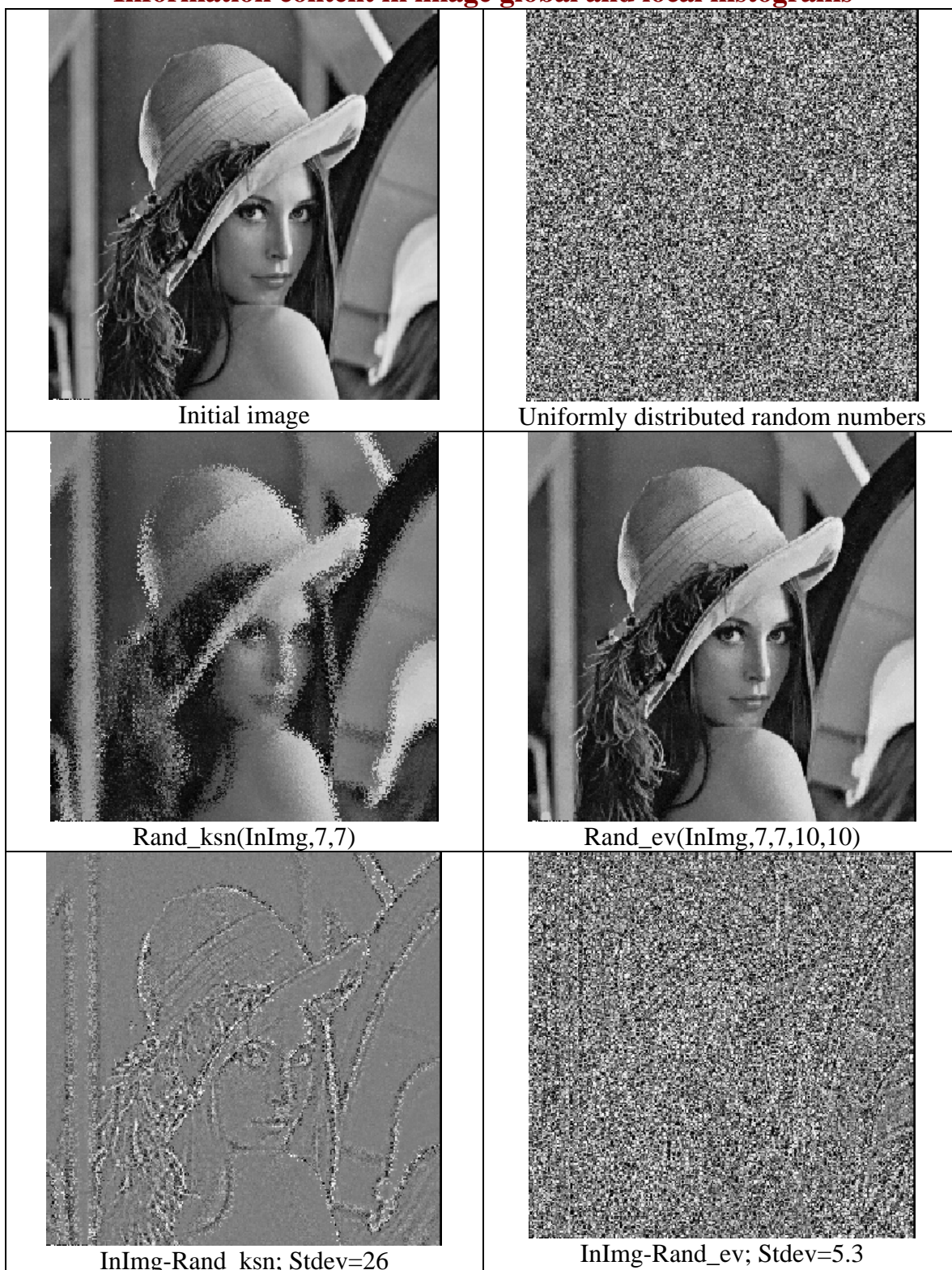
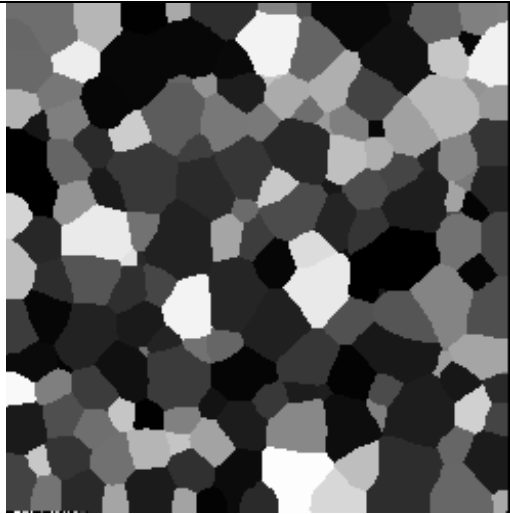
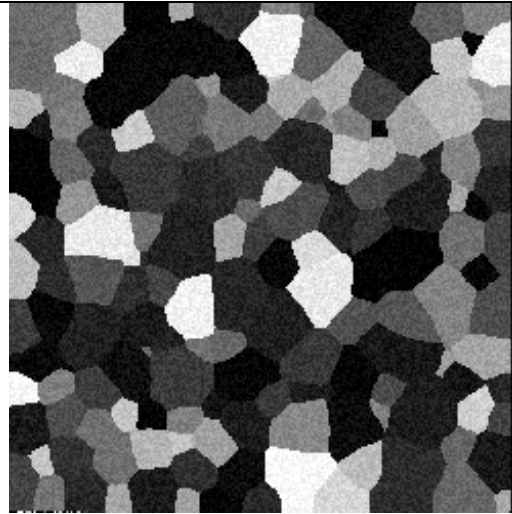


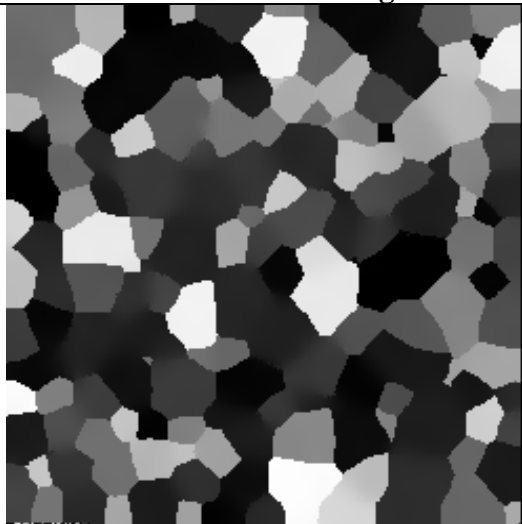
Illustration of information content of local histograms over spatial and EV neighbourhoods. Images in the second row are generated from pseudo-random numbers with the same distribution as that of pixels in spatial window of 7x7 pixels (left) and of pixels in EV-neighbourhood with  $\epsilon V_{plus} = \epsilon V_{minus} = 10$



Noise-free test image

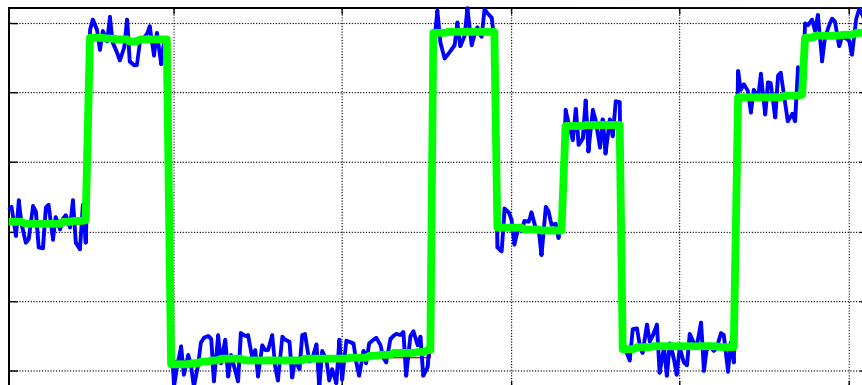
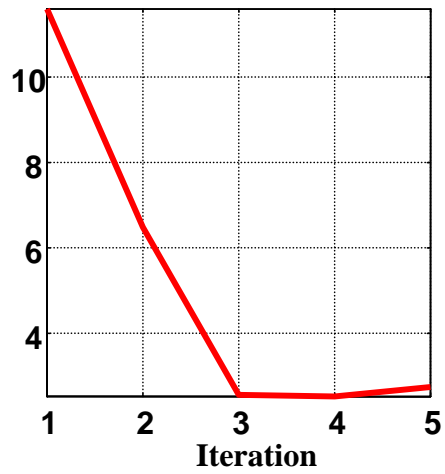


Noisy test image, Noise range =  $\pm 20$

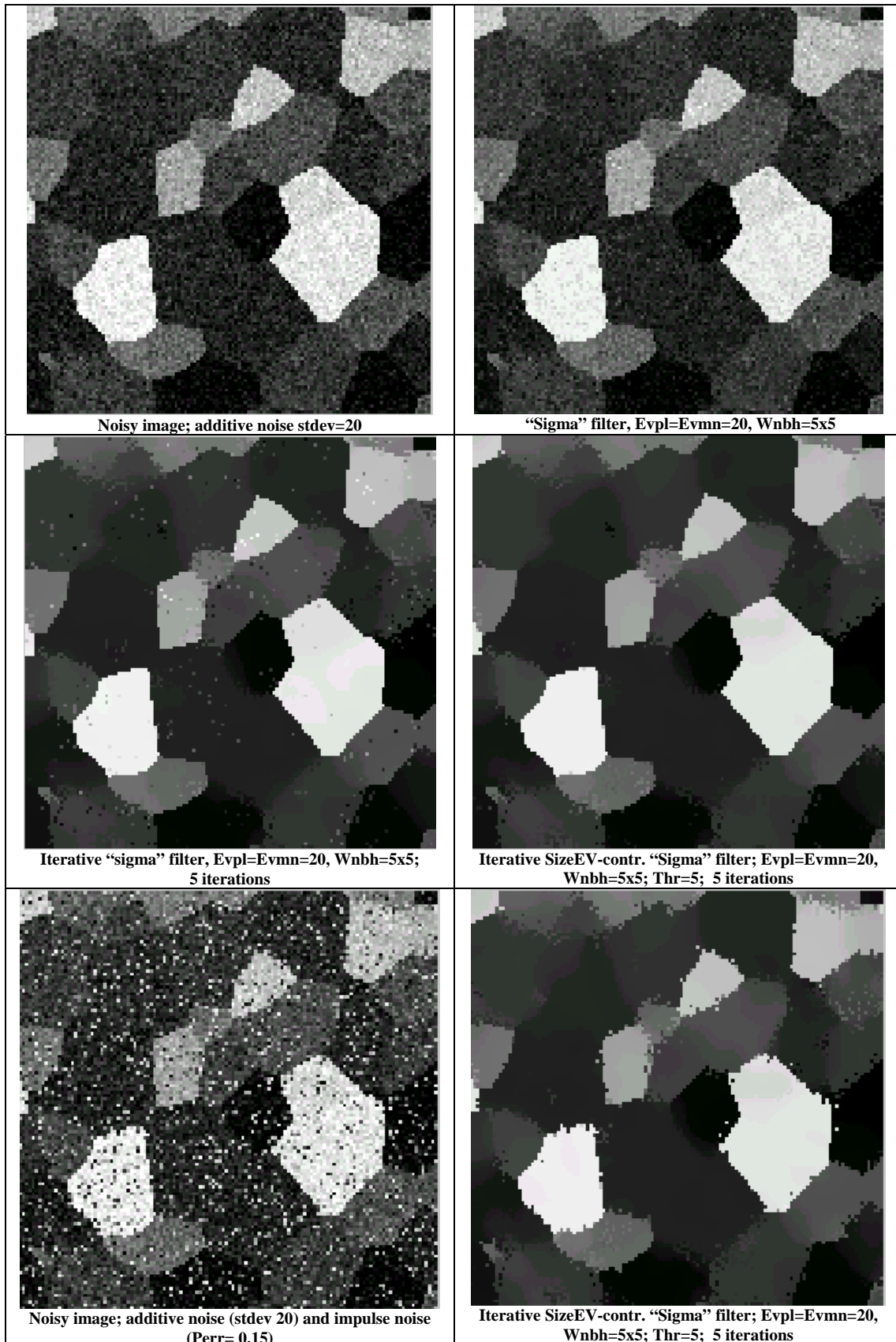


Mean(Evnbh): output image,  
EVpl=20; EVmn=20

Std of residual noise. Init. noise level  $\pm 20$



Plots of 32-th row of the above images (R: original, B: noisy, G: filtered)



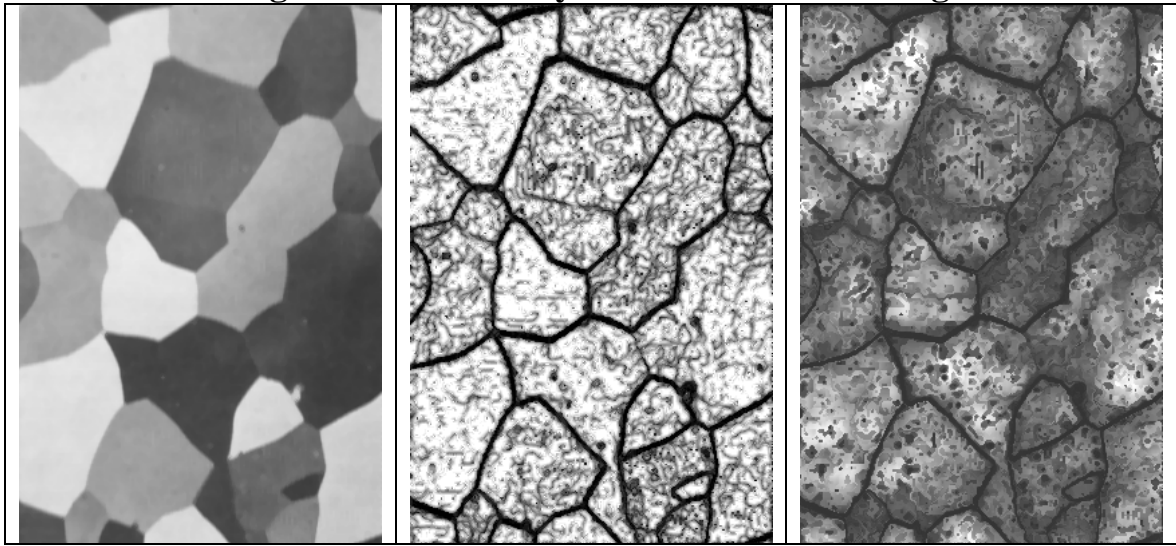
**Fig. 2 Comparison of noise suppression capability of "Sigma"- and Size-EV-controlled "Sigma"-filters**

## Image enhancement by means of rank filtering

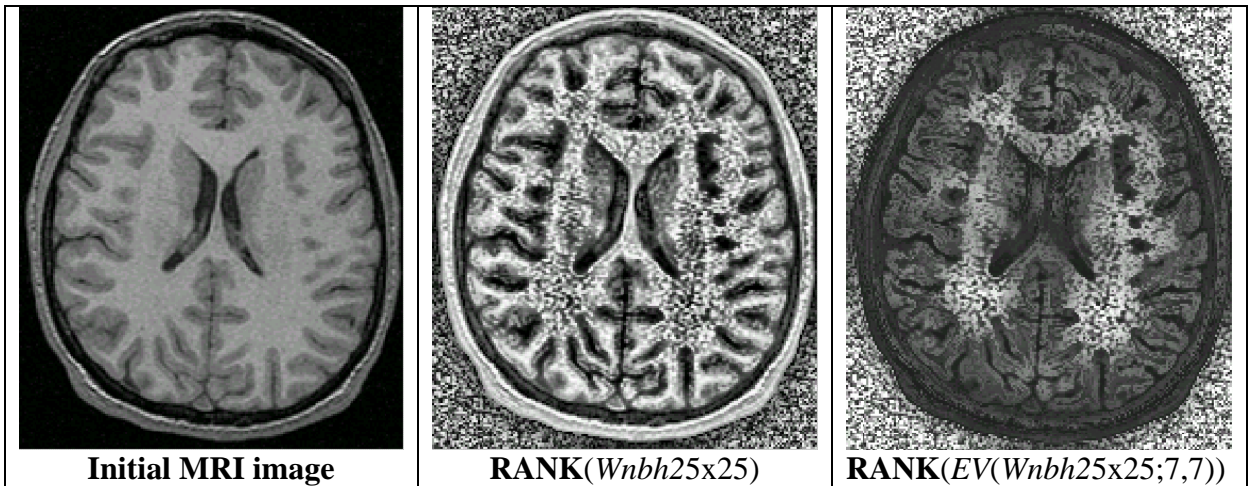


## Component-wise P-histogram equalization

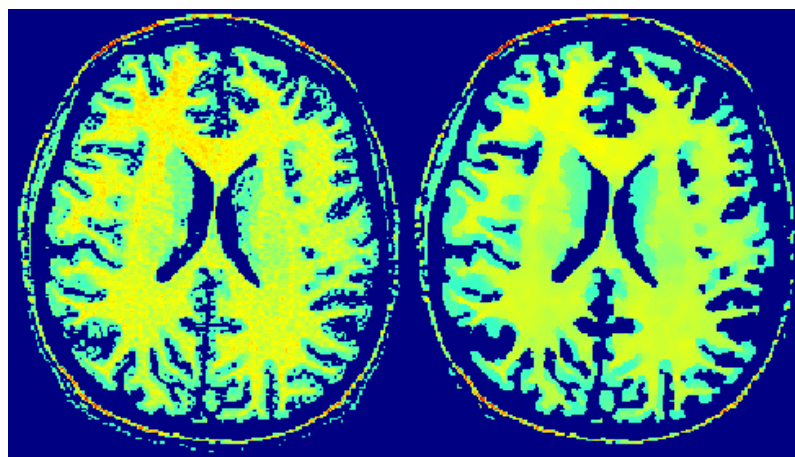
**Image enhancement by means of rank filtering**



**Size-EV and Cardnl-filtering**



**Wnbh and EV-neighborhood local histogram equalization**



**Initial image**

**Segmented image**

`itermnev(InImg,3,3,7,7,5); colormap jet`