

# Reducing image noise

Improving the quality of the electron beam image by means of reducing noise.

Irina Aleksenko

Supervisor: Galina Asova

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# Motivation

The Photo-Injector Test facility at DESY in Zeuthen develops and optimizes **sources for high brightness and short wavelength FELs.**

The main source of information is an image of the transverse distribution of the electron beam obtained with a CCD camera. The main aim for a representative measurement is to get an image that **correctly describes** the electron beam transverse distribution.

Obstacle: **the photodetector signal is combined with noise.**

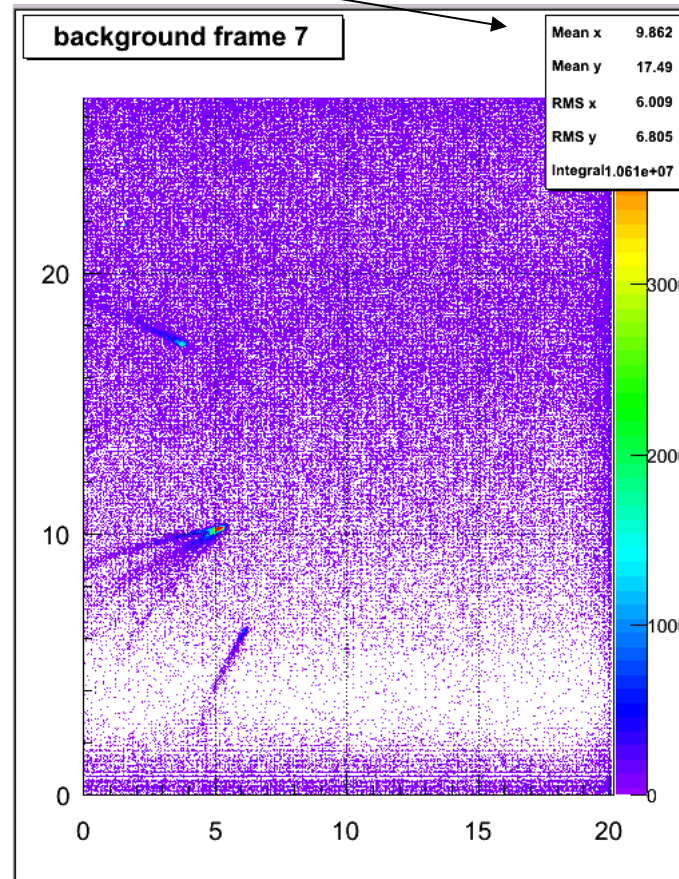
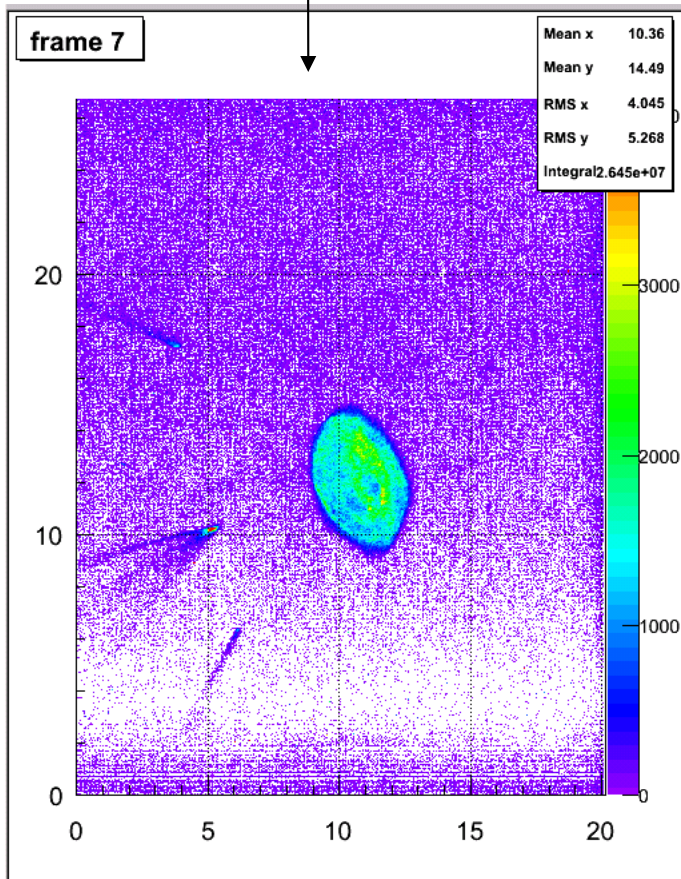
Problem statement: get a high quality image by means of noise reduction.



# PITZ image structure

Result of measurement = {shot<sub>1</sub>, shot<sub>2</sub>, ... shot<sub>n</sub>}

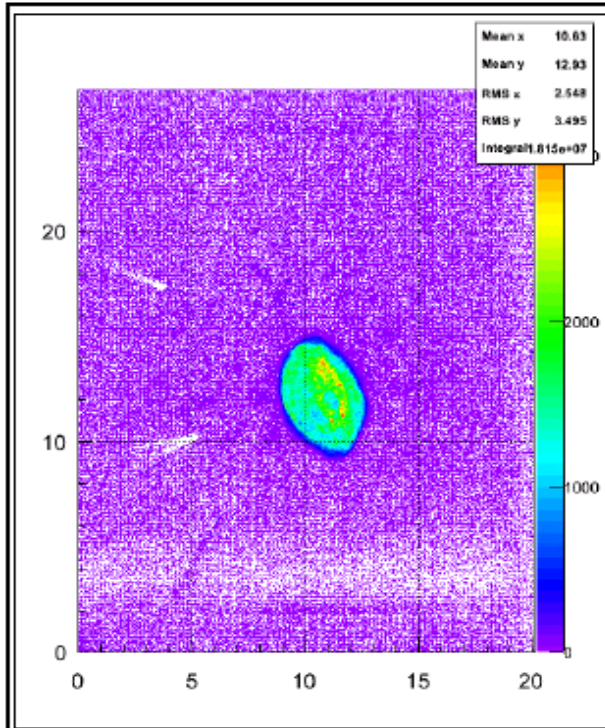
Shot = signal frame + background frame



Noise sources:

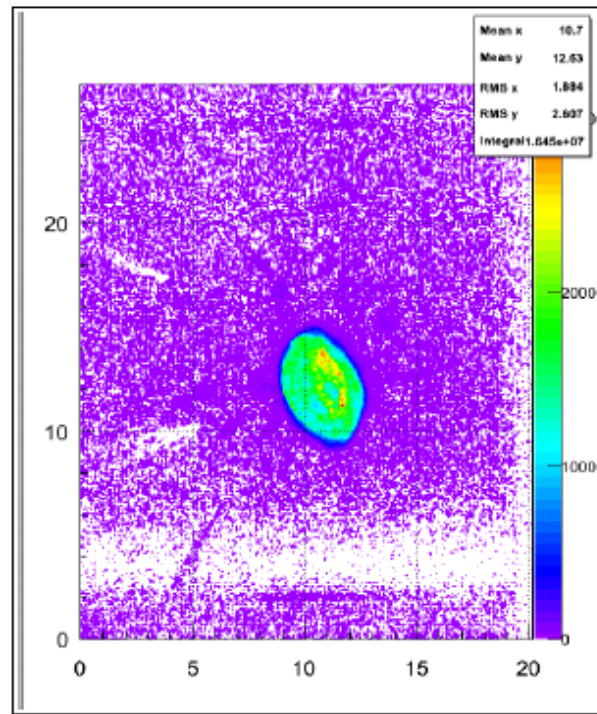
- > dark current
- > CCD enhanced
- > cables
- > electronics
- > X-rays
- > etc.

# Noise. Influence of noise on the measurement precision.



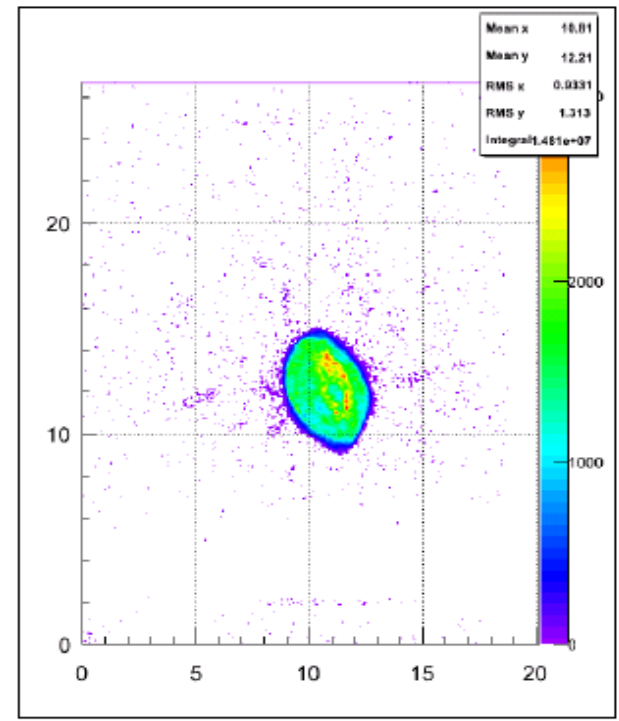
Filter (1)

$$\sigma_x = 2.5 \text{ mm}$$
$$\sigma_y = 3.5 \text{ mm}$$



Filter (2)

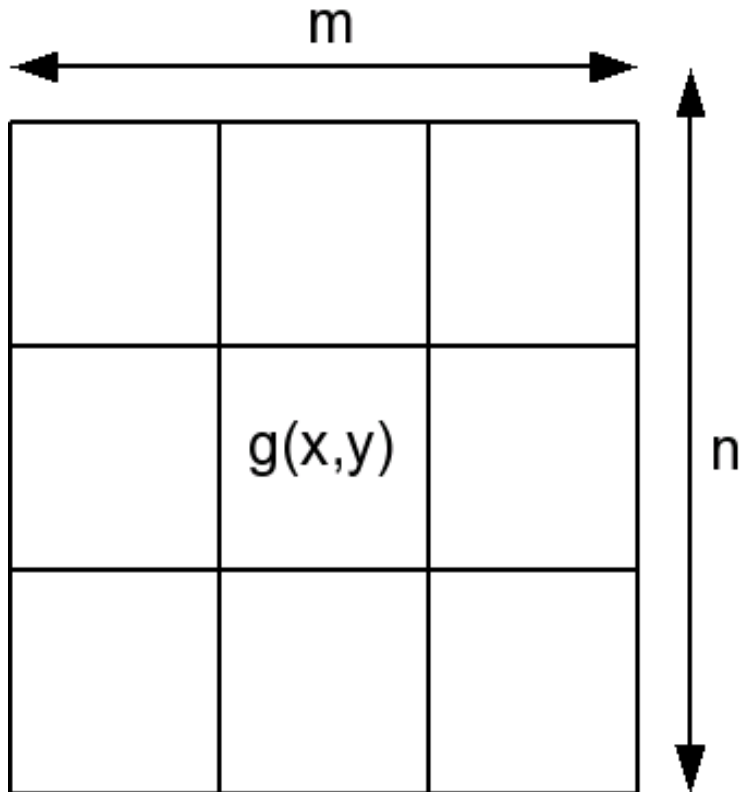
$$\sigma_x = 1.9 \text{ mm}$$
$$\sigma_y = 2.6 \text{ mm}$$



Filter (3)

$$\sigma_x = 0.9 \text{ mm}$$
$$\sigma_y = 1.3 \text{ mm}$$

# Notations



$S_{xy}$  rectangular neighbourhood with the size  $m \cdot n$  and central coordinates  $(x, y)$ .

$g(x,y)$  – measured value in pixel  $(x,y)$

$f(x,y)$  – real wanted value of pixel  $(x,y)$ .

Remark:

$g(x,y) = \langle \text{signal} \rangle - \langle \text{background} \rangle,$

$\langle \text{foo} \rangle$  - mean arithmetic value of pixel foo over the frames

# Filters group

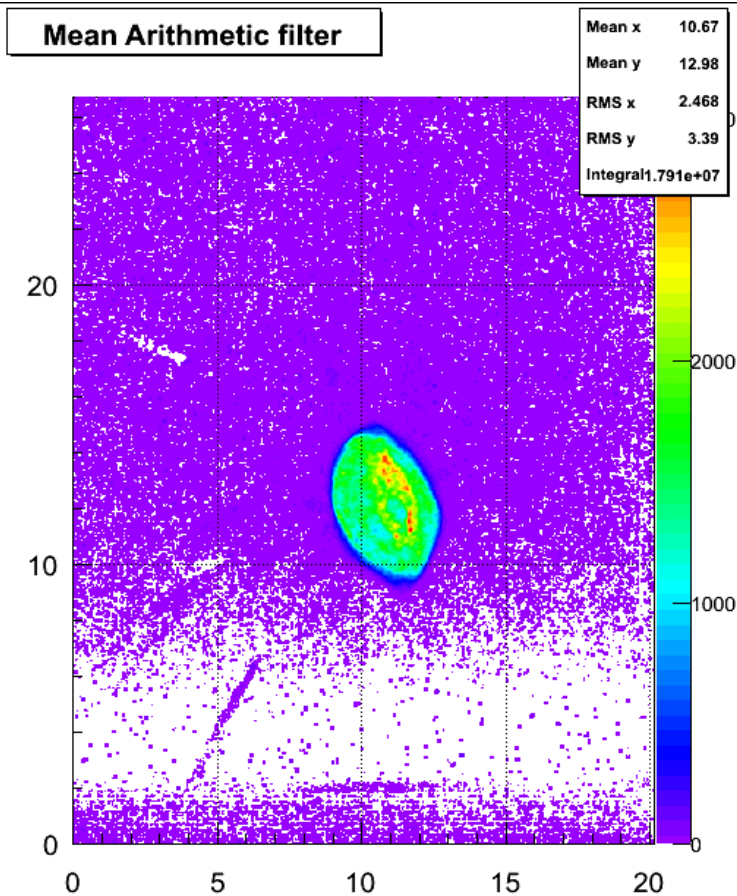
- Averaging filters
- Order statistics filters
- Statistics filters
- Adaptive filters



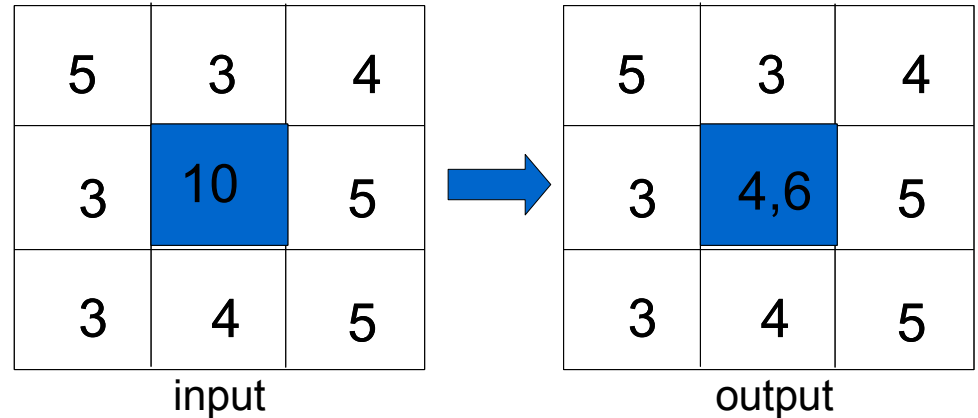
# Averaging filters

Mean Arithmetic filter:

$$f(x, y) = \frac{1}{mn} \sum_{(s, t) \in S_{xy}} g(s, t)$$



$$(5 + 3 + 4 + 3 + 10 + 5 + 3 + 4 + 5) / (3 \times 3) = 4,6$$



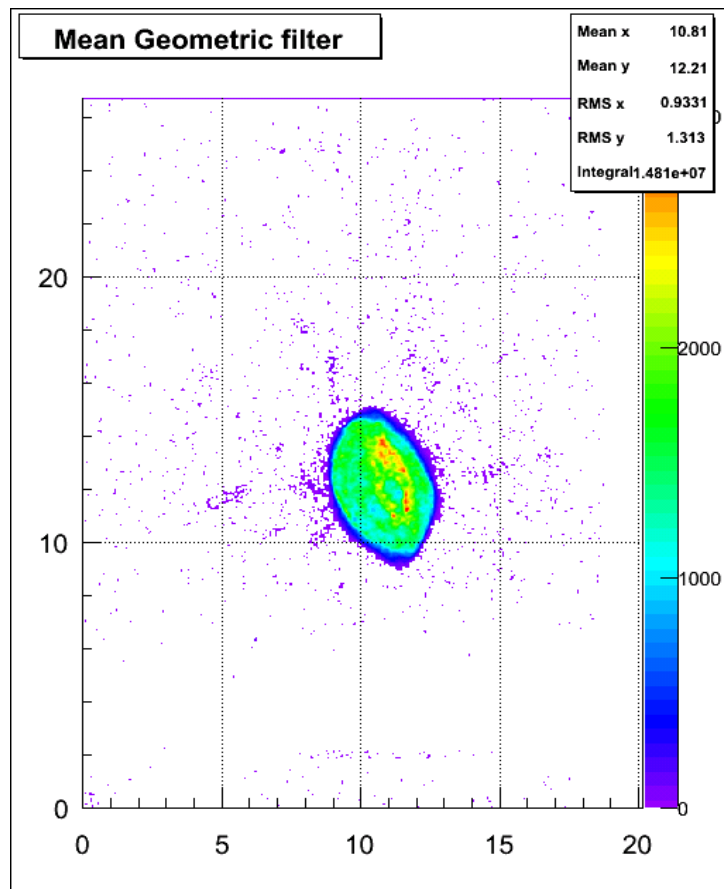
> smooths local variations of the image intensity



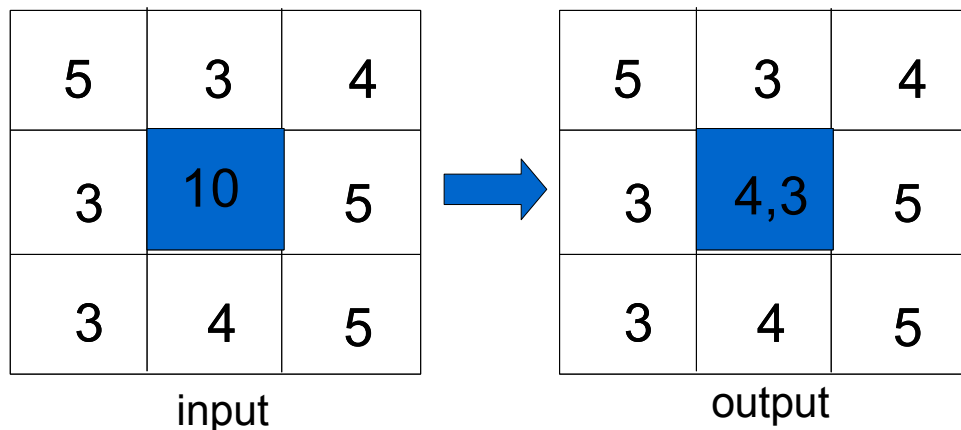
# Averaging filters

Mean Geometric filter:

$$f(x, y) = \left[ \prod_{(s, t) \in S_{xy}} g(s, t)^{\frac{1}{mn}} \right]$$



$$(5 * 3 * 4 * 3 * 10 * 5 * 3 * 4 * 5)^{\frac{1}{3 * 3}}$$



> similar to the Mean Arithmetic filter, but it loses smaller amount of the little details



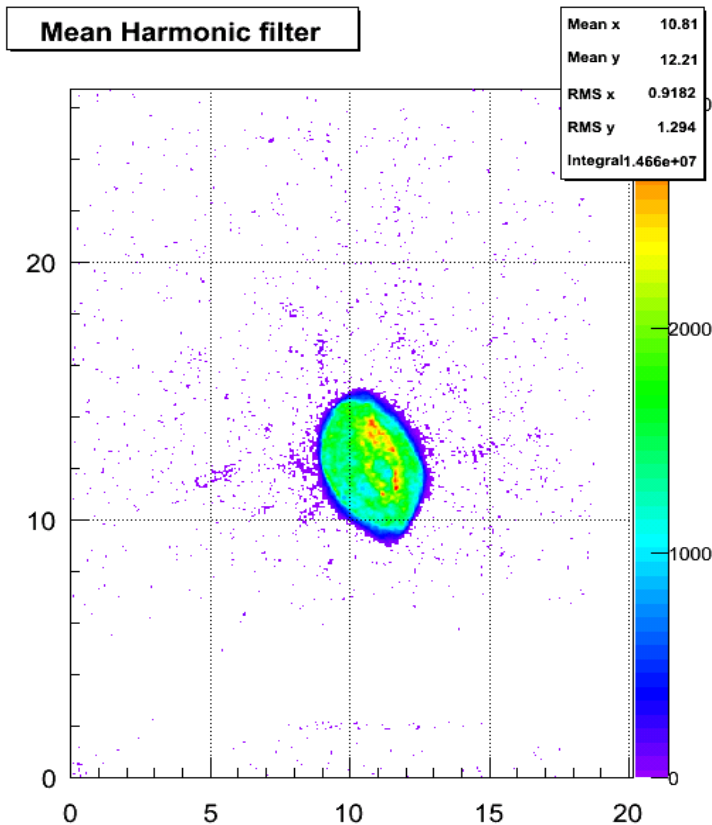
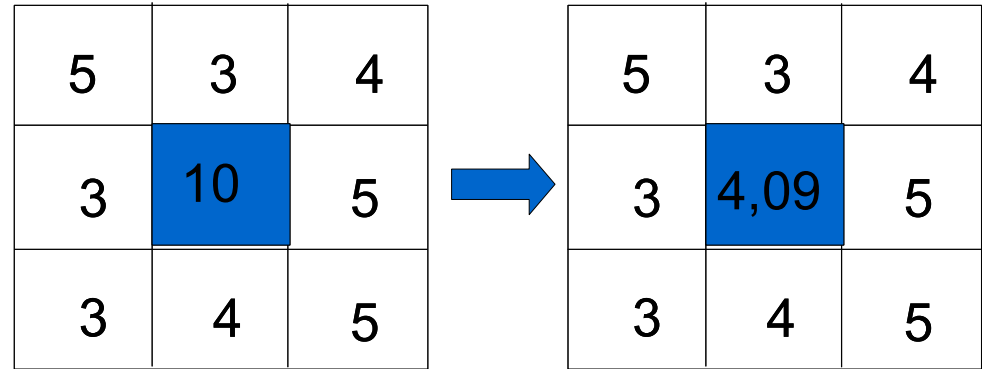


# Averaging filters

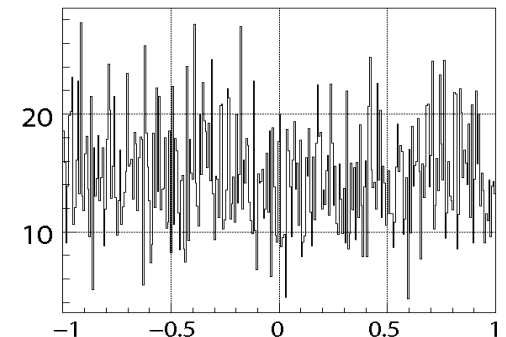
## Mean Harmonic filter:

$$f(x, y) = \frac{mn}{\sum_{(s,t) \in S_{xy}} \frac{1}{g(s,t)}}$$

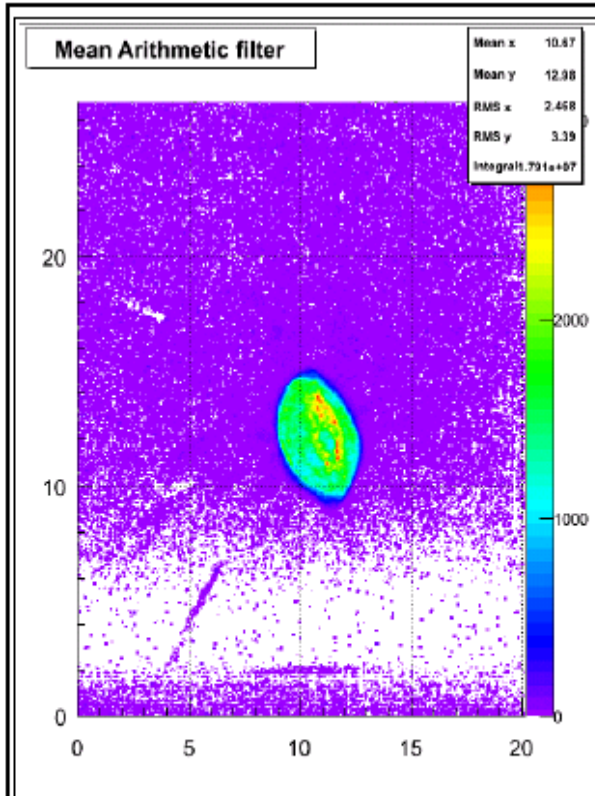
$$3*3 / \left( \frac{1}{5} + \frac{1}{3} + \frac{1}{4} + \frac{1}{3} + \frac{1}{10} + \frac{1}{5} + \frac{1}{3} + \frac{1}{4} + \frac{1}{5} \right) = 4,09$$



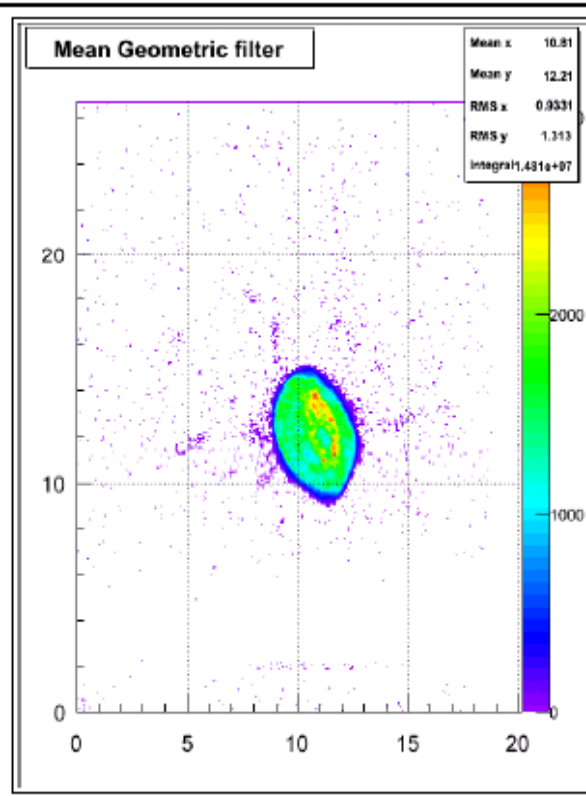
This filter is useful in the case of the white Gaussian noise



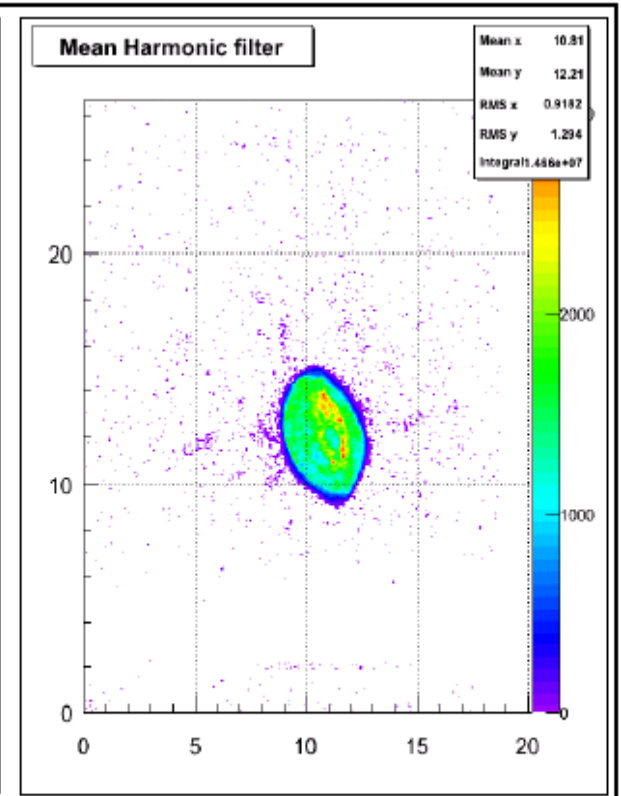
# Examples of averaging filters



Mean Arithmetic

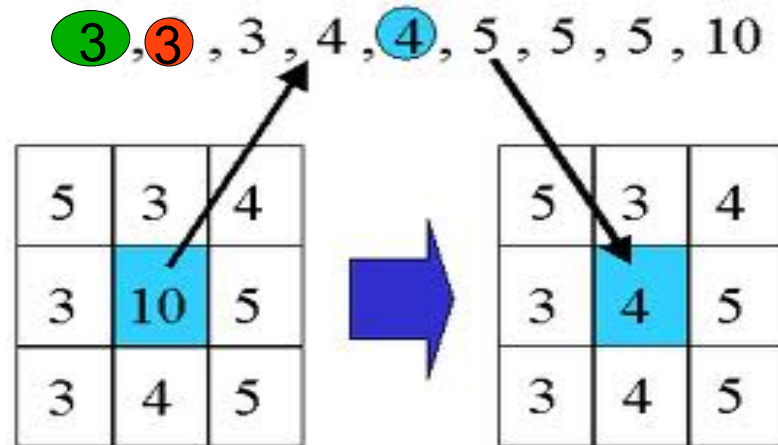


Mean Geometric



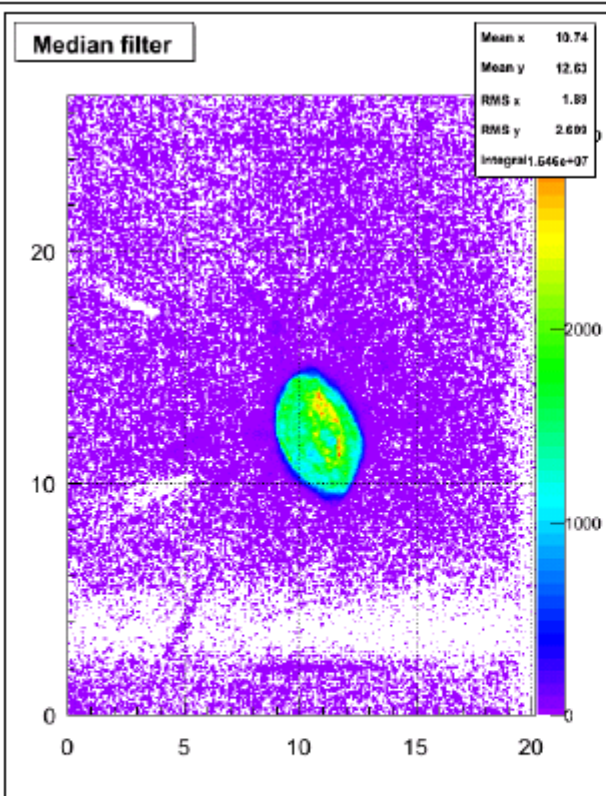
Mean Harmonic

# Order statistic filters

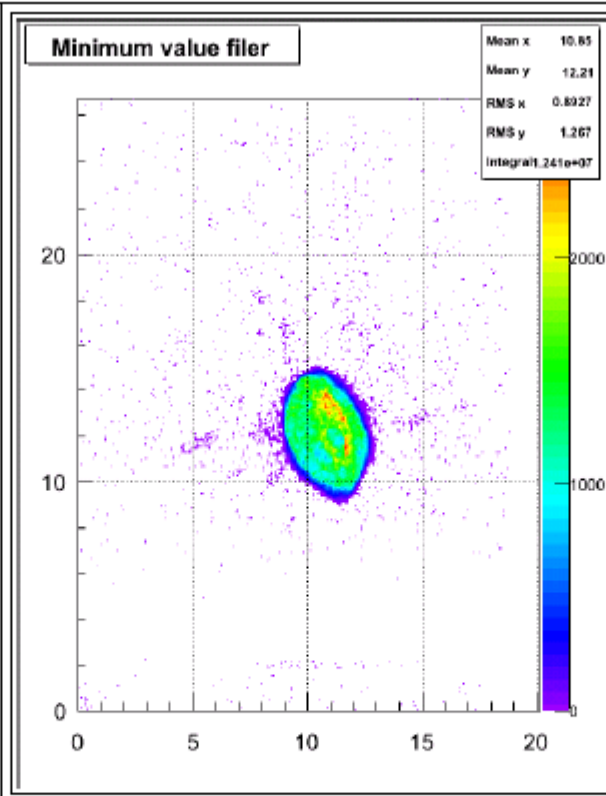


- **Median filter:**  $f(x, y) = \text{med}\{g(s, t) : (s, t) \in S_{xy}\}$
- **Minimum value filter:**  $f(x, y) = \min\{g(s, t) : (s, t) \in S_{xy}\}$
- **Modified minimum value filter:**  $f(x, y) = \text{next after } \min\{g(s, t) : (s, t) \in S_{xy}\}$

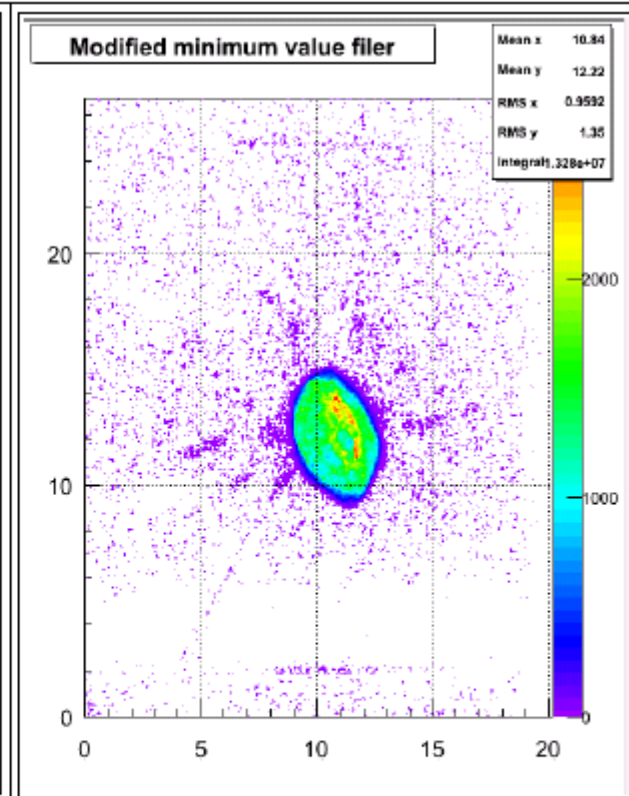
# Example of Order statistic filters



Median Filter



Minimum Value



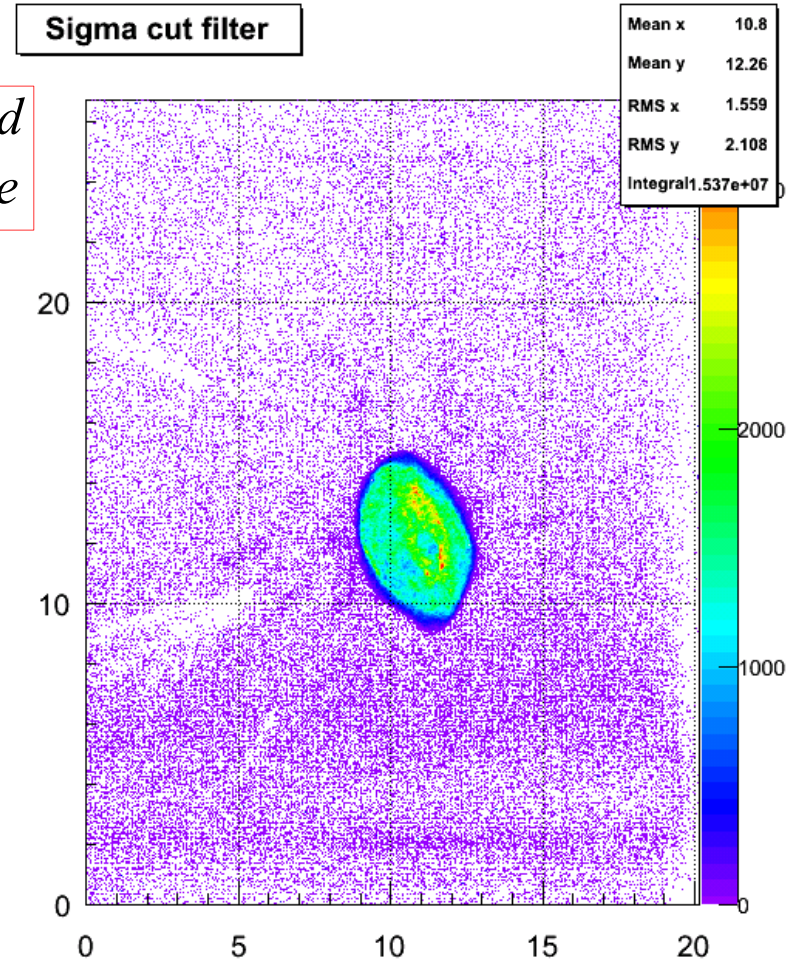
Modified Minimum Value

# Statistics filters

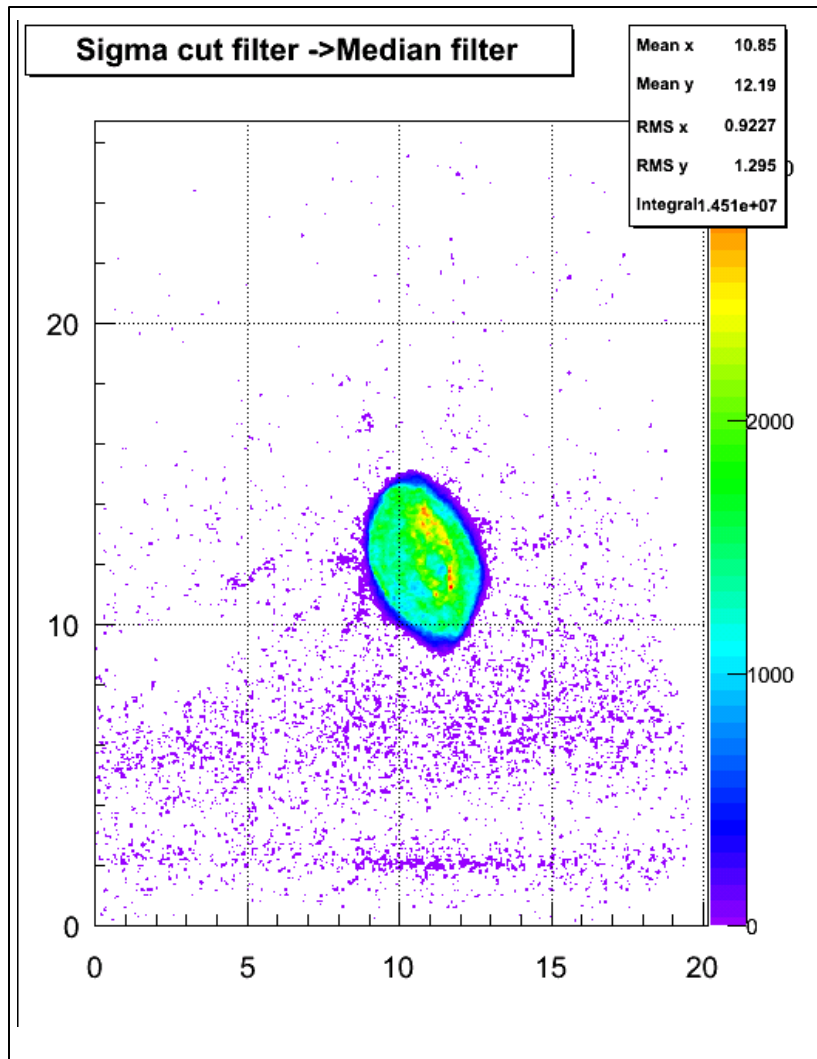
Sigma cut filter:

$$f(x, y) = \begin{cases} 0, & (g(x, y) - \sigma_{noise}(x, y)) < Threshold \\ g(x, y) & \text{else} \end{cases}$$

This filter is useful when the signal pixel intensity is much bigger than one sigma of the noise pixel intensity. Its idea is that **high Intensity signal is regarded as the electron beam signal, low signal - as noise.**



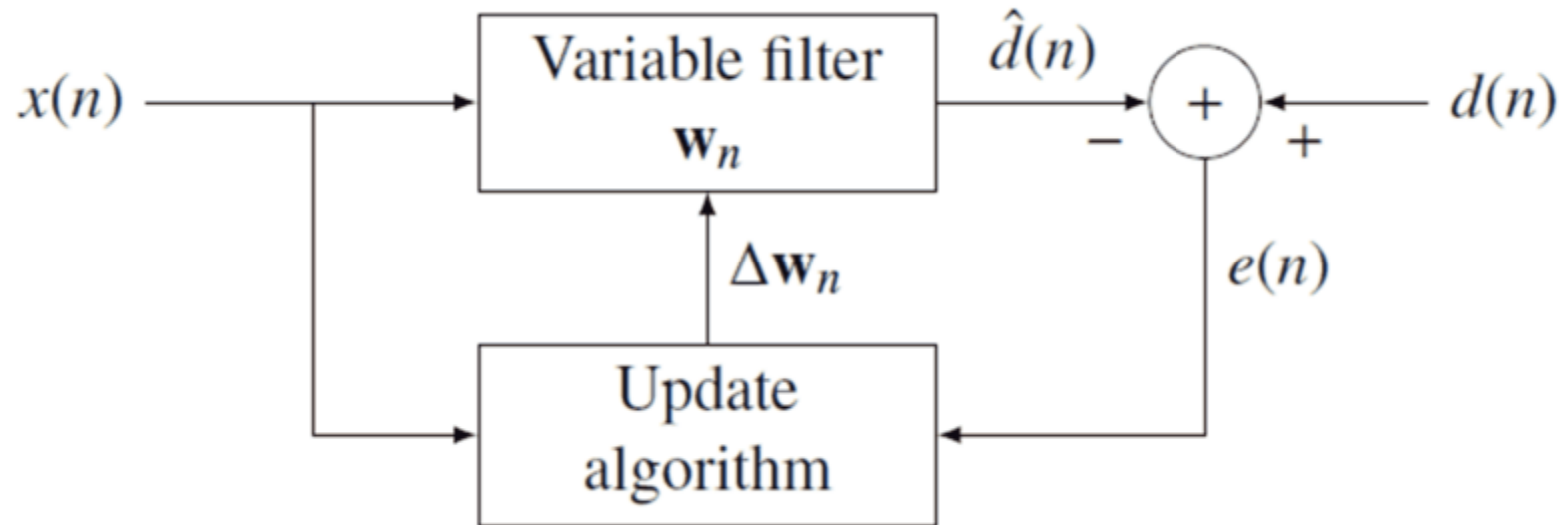
# Example of filter combination



Sigma cut filters is a good filter for preparing the data image before applying another filter. It rarefies noise well. Afterwards filters, which work with big amount of the noise unsuccessful, demonstrates enough good result.

# Adaptive filters

An adaptive filter is a filter that self-adjusts its transfer function according to an optimizing algorithm.



Wiener filter, Least Mean Squares filter





# Conclusion

Result:

- > Several filters were applied, none of them delivers satisfactory result
  
- > Future work:
  - apply different combination of filters
  - apply filters to each frame and then calculate average image
  - analyze adaptive filtering, especially LMS filter



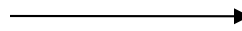
Thank you for your attention



# A priori noise information

Types of noise:

➤ “salt and pepper” noise



➤ Gaussian noise:

$$p(z) = \frac{1}{\sqrt{2\pi}\sigma} * e^{-\frac{(z-\mu)^2}{2\sigma^2}},$$

where  $p(z)$  is probability density function,  $\mu$  and  $\sigma^2$  are the mean and the variance.



# Denoising

Denoising (noise reduction, noise reducing) is the extraction of a signal from a mixture of signal and noise.

Filters with finite impulse response (FIR) . FIR is a property of signal processing systems. FIR filter is a type of a discrete-time filter. The impulse response, the filter's response to a Kronecker delta input, is finite because it settles to zero in a finite number of sample intervals.

Filters with infinite impulse response (IIR). IIR systems have an impulse response function that is non-zero over an infinite length of time.

Digital image -> finite impulse response

