



— VSNR —

# Variational Stationary Noise Remover

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Third light sheet microscopy workshop



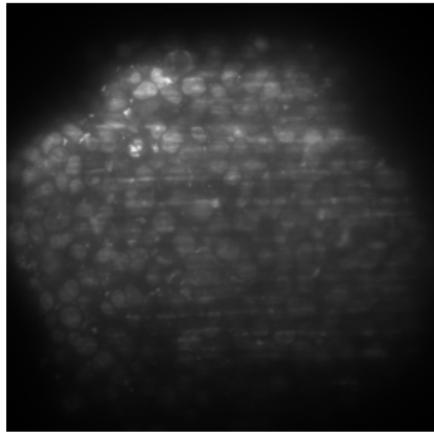
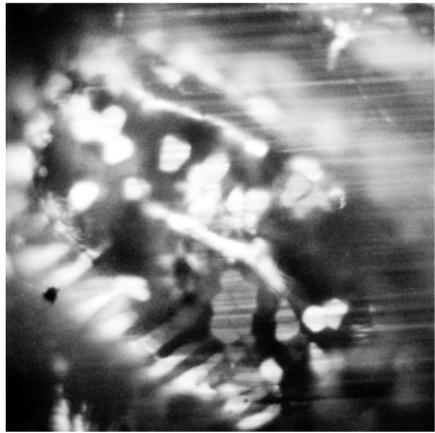
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## Starting observation

In many applications, **structured noise** degrades the images.

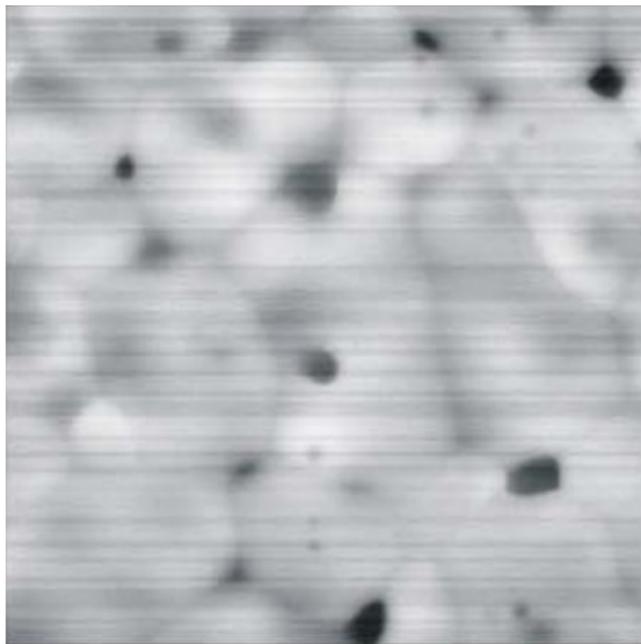


**SPIM** - Stripes due to light absorption and scattering.

Left: Xenopus laevis's late tailbud (40X NA 0.8).

Right: Multicellular Tumor Spheroid (20X NA 0.5).

## Examples of images with stripes

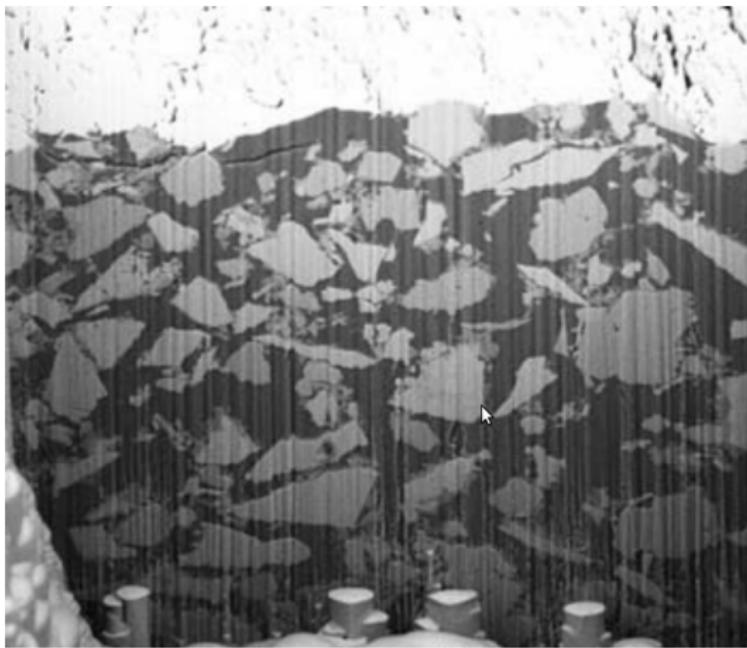


Scanning electron microscope: Stripes in a sintered specimen of Cerium Oxyde.

[Chen et al] *DeStripe: frequency-based algorithm for removing stripe noises from AFM images*. BMC Structural

Biology 2011.

## Examples of images with stripes



**Ion beam nanotomography:** Stripes in particles of cement paste.

[Münch et al] *Stripe and ring artifact removal with combined wavelet - Fourier filtering*. Optical express 2009.

## Other applications where correlated noise occurs.

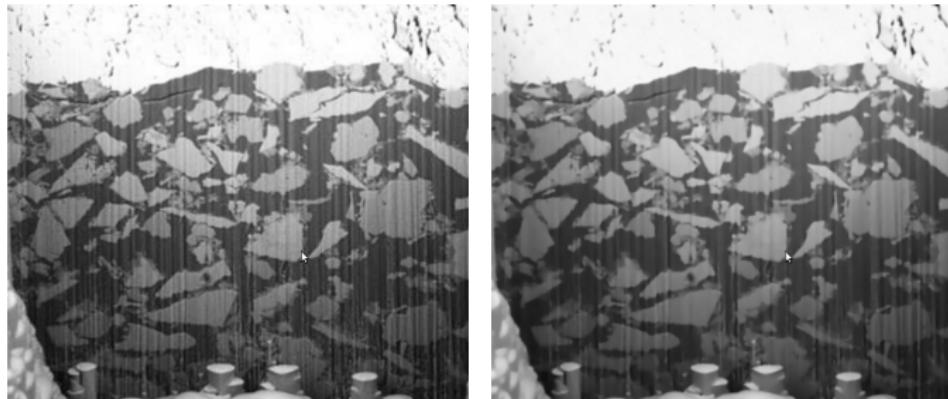
- SPIM.
- Atomic force microscopy.
- Electron tomography.
- Synchrotron X-ray microscope.
- Ion beam nanotomography (waterfall effect).
- MODIS images (atmosphere imaging).
- Digital elevation models (satellite imaging).
- Imaging under turbulence.
- ...

## Motivation - Standard denoising methods fail



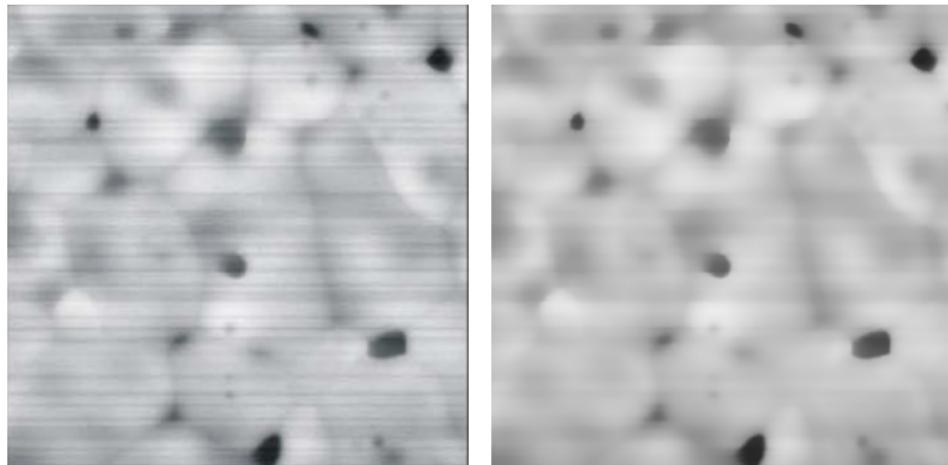
Left: original image. Right: denoised image using Gaussian smoothing.

## Motivation - Standard denoising methods fail



Left: original image. Right: denoised image using anisotropic diffusion.

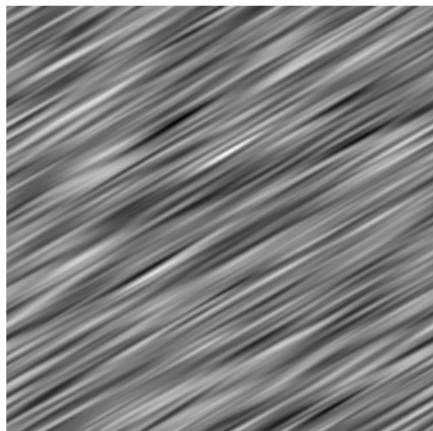
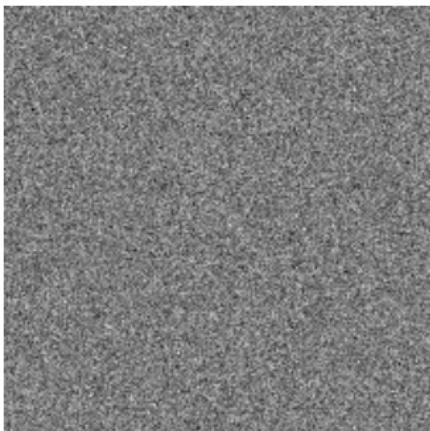
## Motivation - Standard denoising methods fail



Left: original image. Right: denoised image using bilateral filtering.

# Why do standard methods fail ?

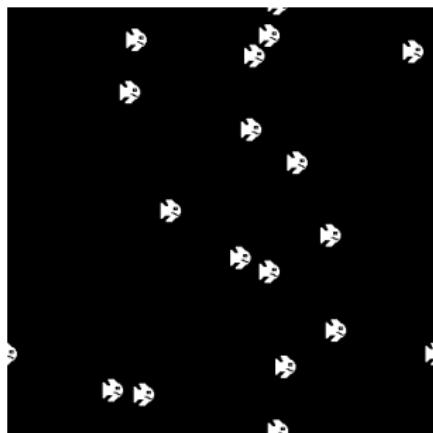
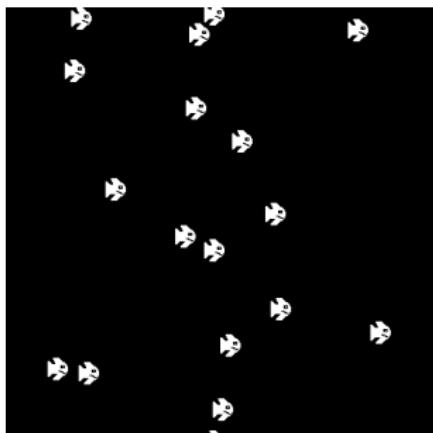
- **Main reason:** standard methods rely on a **white noise** assumption. White means uncorrelated pixelwise.
- **Our objective:** design methods for **correlated/stationary** noise.



White noise (left) VS stationary noise (right) .

# What is a stationary noise ?

Translating the sample in space **does not change its probability.**



Left: A sample of stationary noise.

Right: the same sample translated.

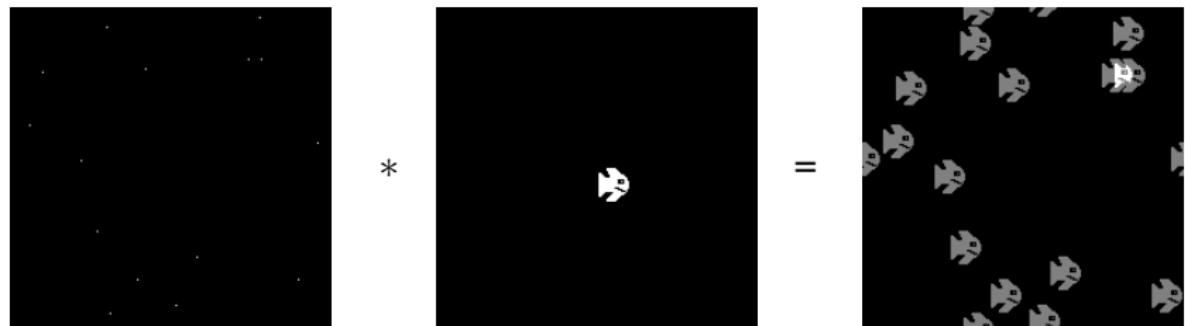
**A natural assumption:** we have **no *a priori*** knowledge on the location of features. They appear randomly.

# How can we generate stationary noises ?

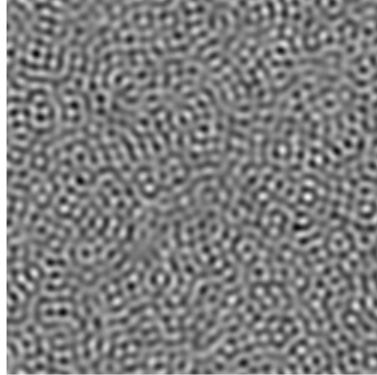
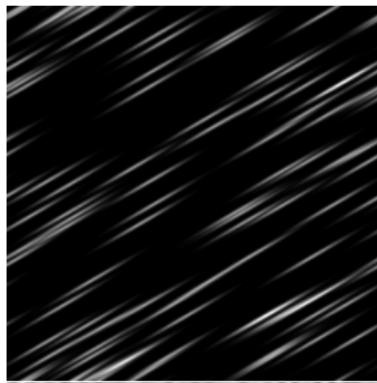
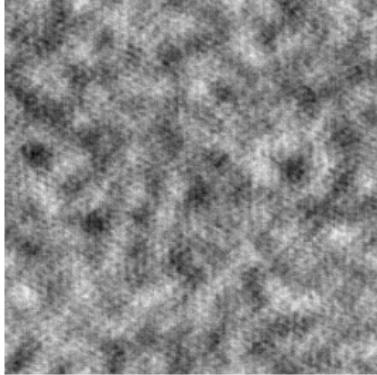
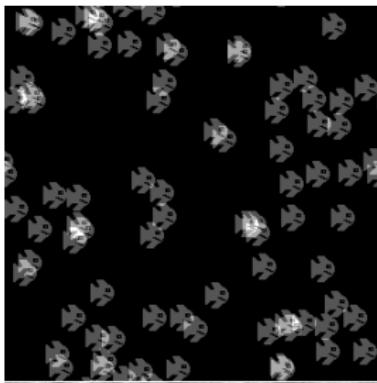
- The class of stationary noises is **too wide** for numerical processing.
- We restrict to the class of noises obtained by  
**Replicating and translating an elementary pattern  $\psi$ .**

This can be achieved by convolving white noise with a **pattern**:

$$\lambda * \psi(\mathbf{x}) = \sum_{\mathbf{y}} \lambda(\mathbf{y}) \psi(\mathbf{x} - \mathbf{y}).$$



## Examples of stationary noises



# Model of image formation

A noisy image  $u_0$  is the sum of:

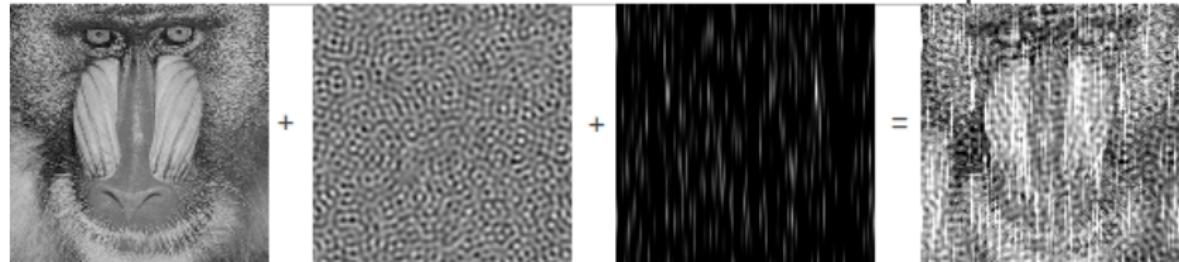
- the original image  $u$ .
- a stationary noise  $b$ .

$$u_0 = u + b$$

where

$$b = \sum_{i=1}^m \lambda_i * \psi_i$$

$b$  is a sum of stationary processes.



# The VSNR algorithm

## INPUT:

➤ *A pattern:*



➤ *A white noise statistics:*



➤ *A regularization parameter:* tunes the algorithm.

## OUTPUT:

➤ *A “nice” image.*

# The VSNR algorithm

The algorithm finds the most likely image.

Turns out to be a **convex optimization problem**.

$$\underset{\lambda \in \mathcal{R}^{m \times n}}{\operatorname{argmin}} \left( \left\| \nabla \left( u_0 - \sum_{i=1}^m \lambda_i * \Psi_i \right) \right\|_1 + \sum_{i=1}^m \phi_i(\lambda_i) \right)$$

More details in :

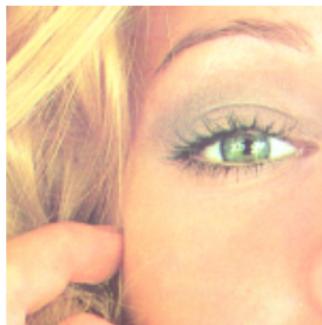
[Fehrenbach et al] *Variational algorithms to remove stationary noise. Application to SPIM imaging*. Preprint 2011.

[Fehrenbach et al] *Variational algorithms to remove stripes: a generalization of the negative norm models*. 2011.

## Examples of application - simulated data (1)

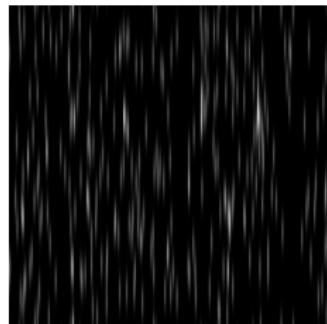
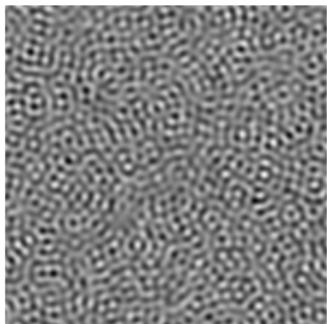
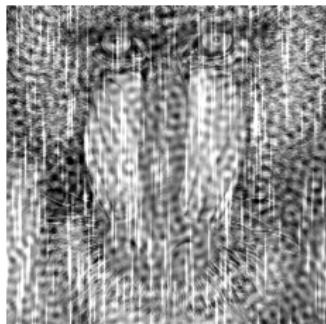


Left: noisy image. Right: detail.

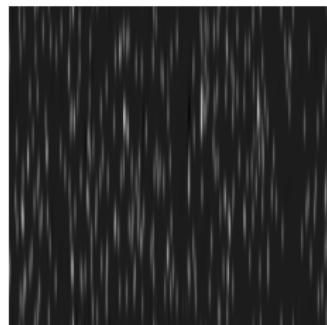
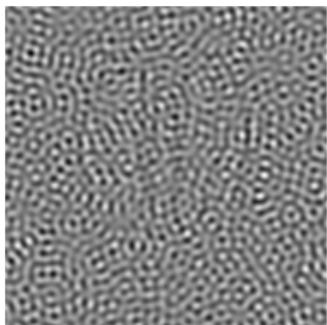


Denoised images.

## Examples of application - simulated data (2)

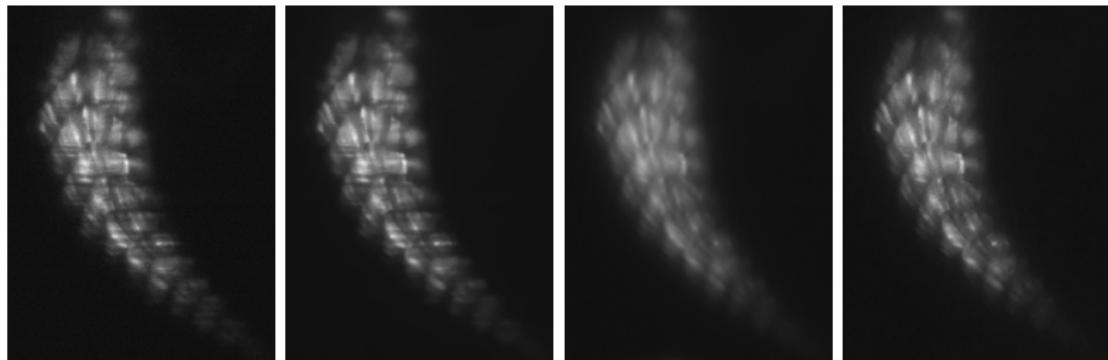


Left: noisy image. Mid: 1st component. Right: 2nd component.



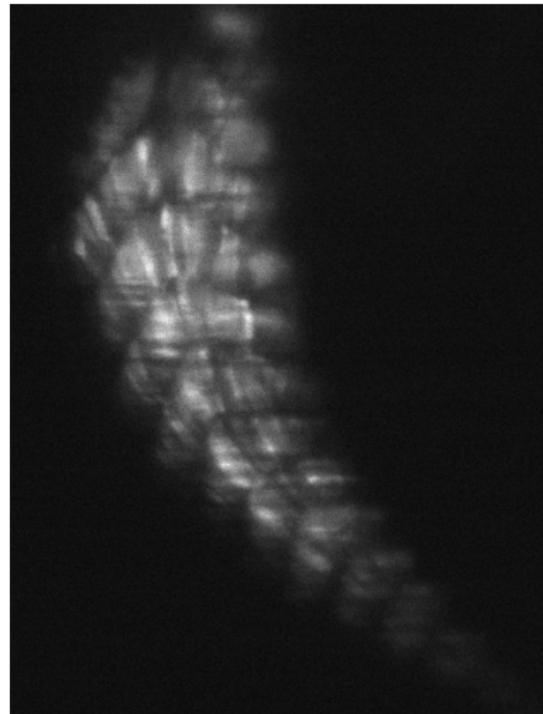
Recovered components.

## Examples of application - SPIM image of a zebrafish



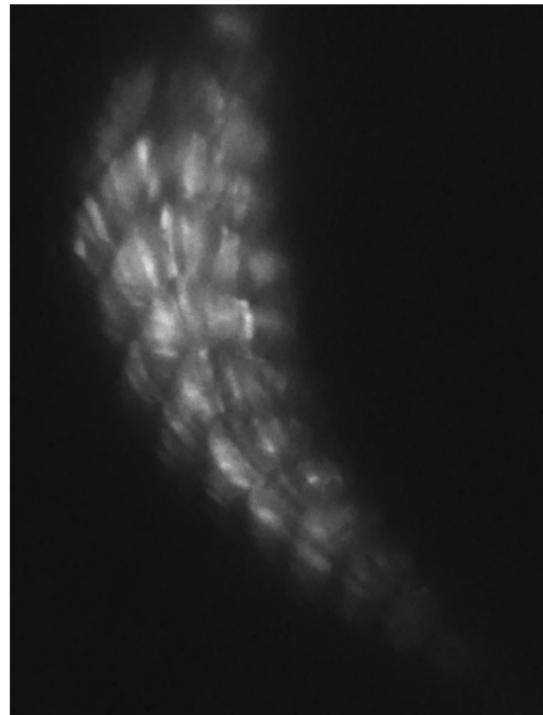
Original - TV-L2 (standard) -  $H^1$ -norm (fast) - VSNR

## Examples of application - SPIM image of a zebrafish



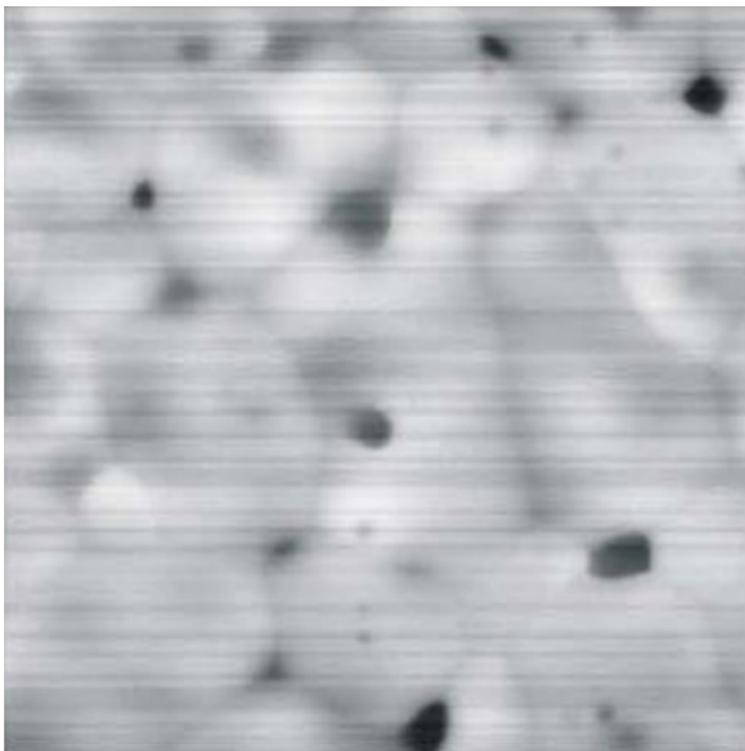
Original

## Examples of application - SPIM image of a zebrafish



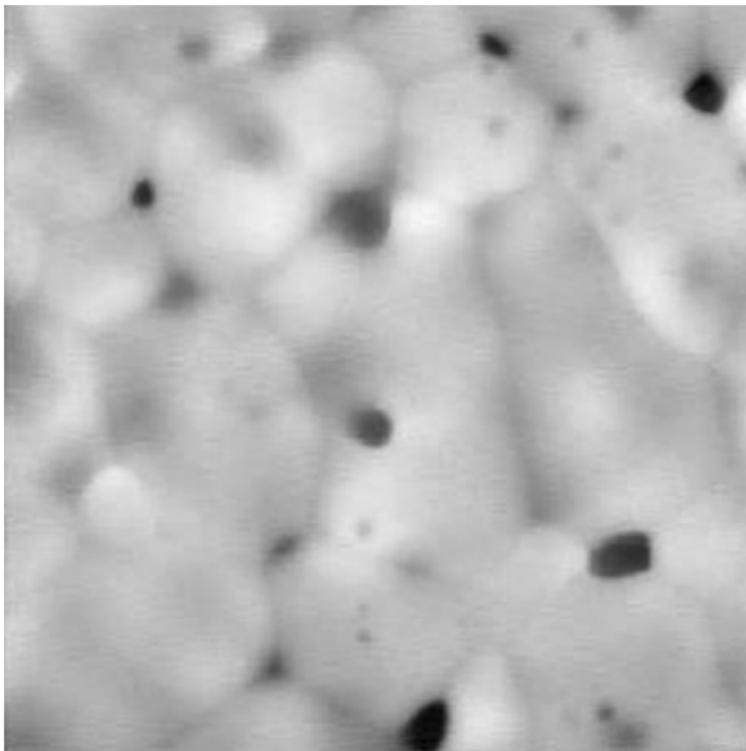
Denoised

## Examples of application - Ion beam nanotomography



Original

## Examples of application - Ion beam nanotomography



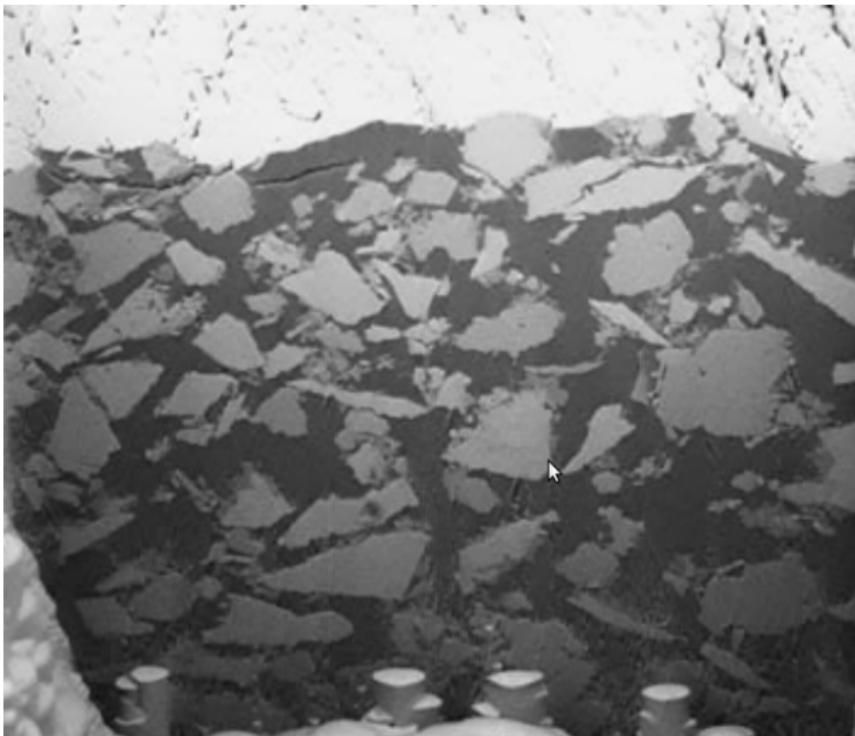
Denoised

## Examples of application - SEM



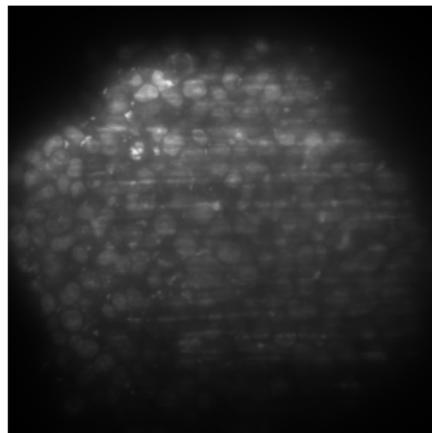
Original

## Examples of application - SEM

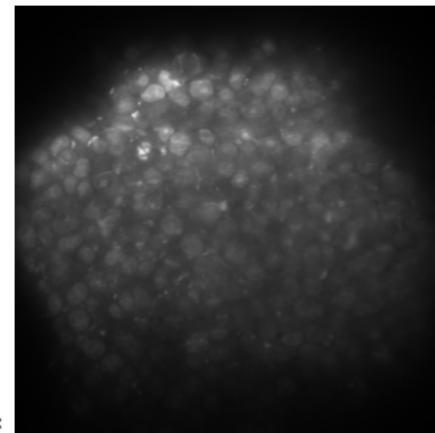


Denoised

## Examples of application - SPIM image of a spheroid



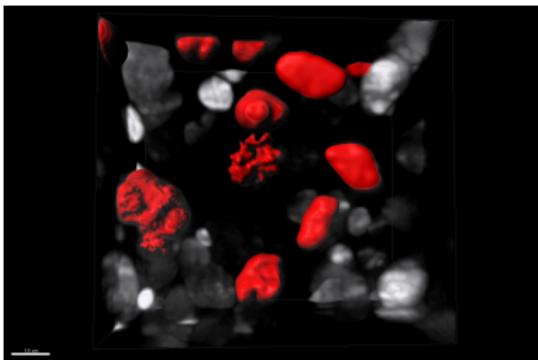
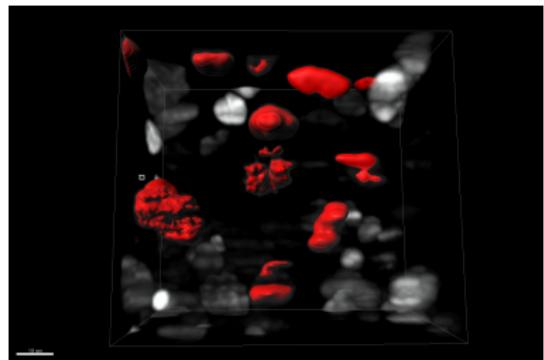
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## Examples of application - SPIM image



3D rendering using Imaris.  
Left: original. Right: denoised.

# Conclusion

## Main messages:

- Standard methods unadapted to the removal of **correlated noise**.
- Development of a versatile method for **stationary noise**.
- New theoretical bases (see preprints).
- Matlab implementation available on :  
[www.math.univ-toulouse.fr/~weiss/index.html](http://www.math.univ-toulouse.fr/~weiss/index.html)

## Perspectives:

- Real 3D implementation.
- Acceleration using GPU programming.
- FIJI implementation.

[Fehrenbach et al] *Variational algorithms to remove stationary noise. Application to SPIM imaging*. Preprint 2011.

[Fehrenbach et al] *Variational algorithms to remove stripes: a generalization of the negative norm models*. 2011.

## Ending words

- ♣ Thanks you for your attention ♣ -

– ♥ Thanks again to the organizers ♥ –

— ♠ Please welcome warmly the next magical speaker! ♠ —