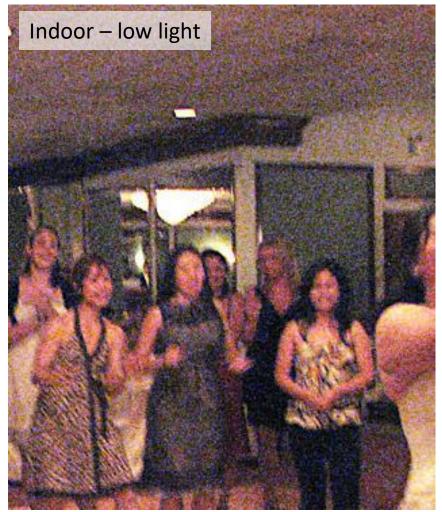
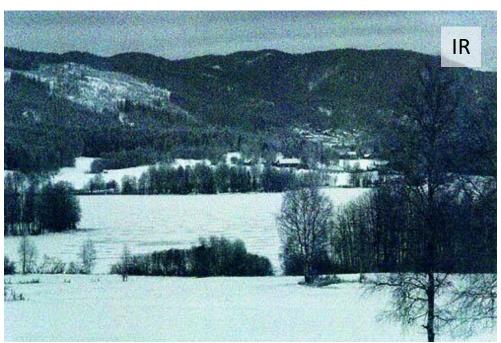
Estimation and removal of noise from Images

Noise

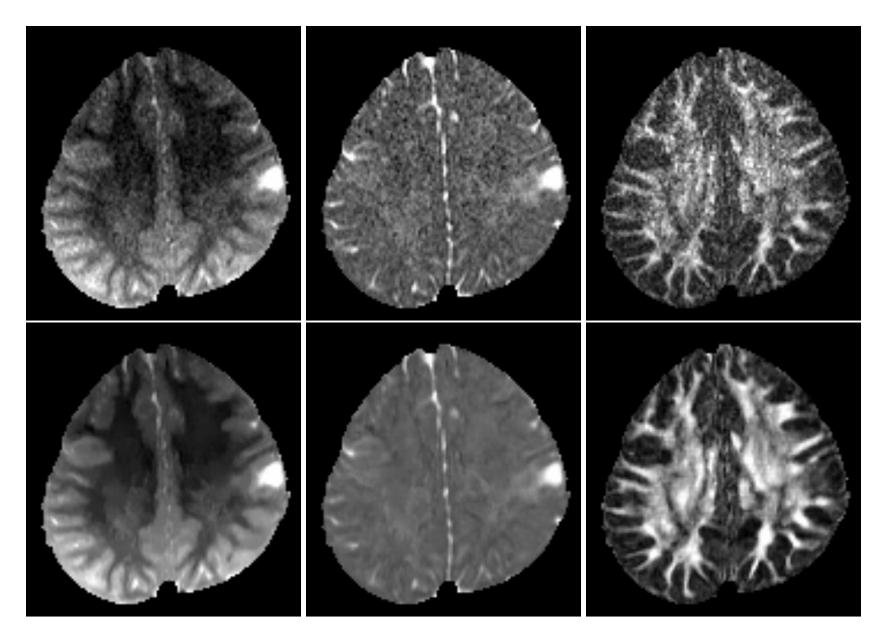
- Where does noise come from?
 - Sensor (e.g., thermal or electrical interference)
 - Environmental conditions (rain, snow etc.)
- Why do we want to denoise?
 - Visually unpleasant
 - Bad for compression
 - Bad for analysis







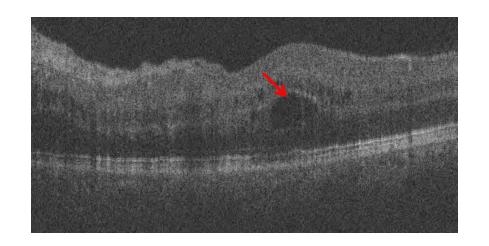


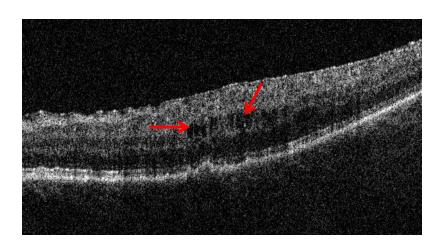


Diffusion weighted images: Before and after denoising

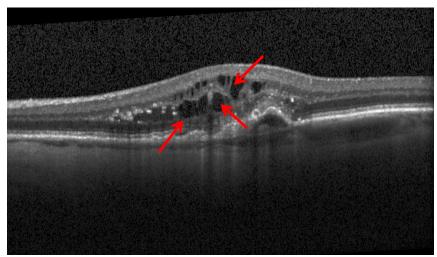


Challenges: Scan quality and Vendor differences



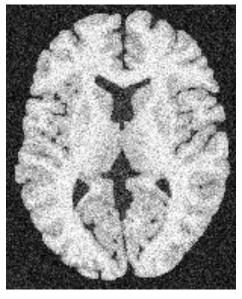


Cirrus Nidek





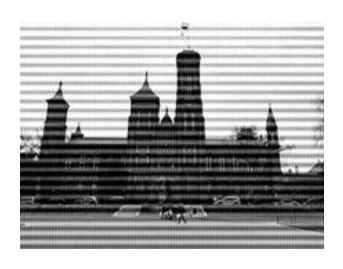
Spectralis Topcon







Salt & pepper



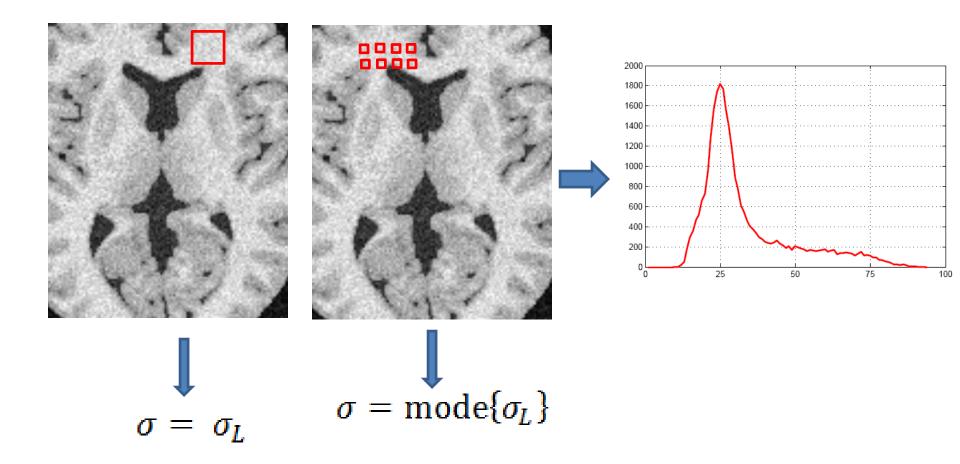
Periodic noise

Which image do you prefer?

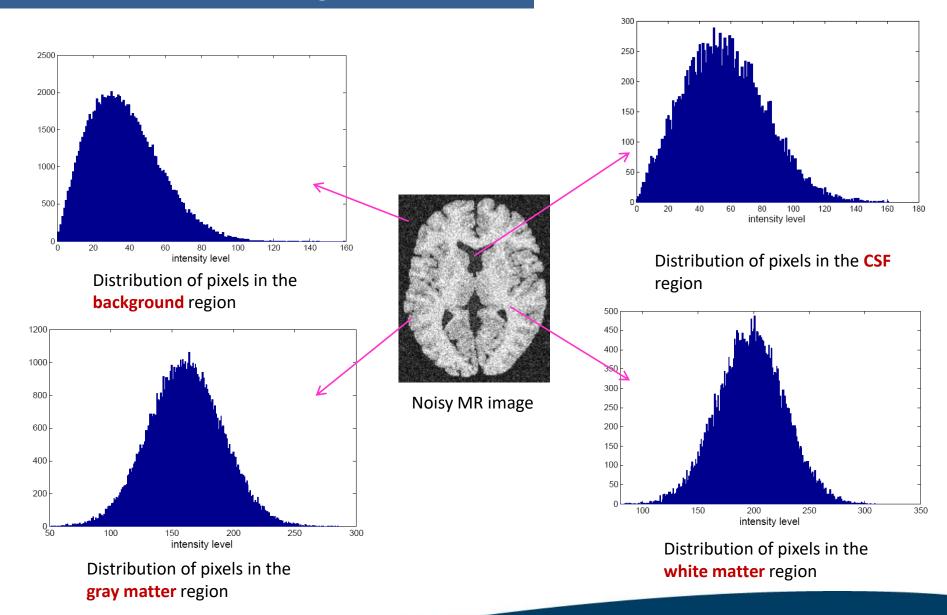




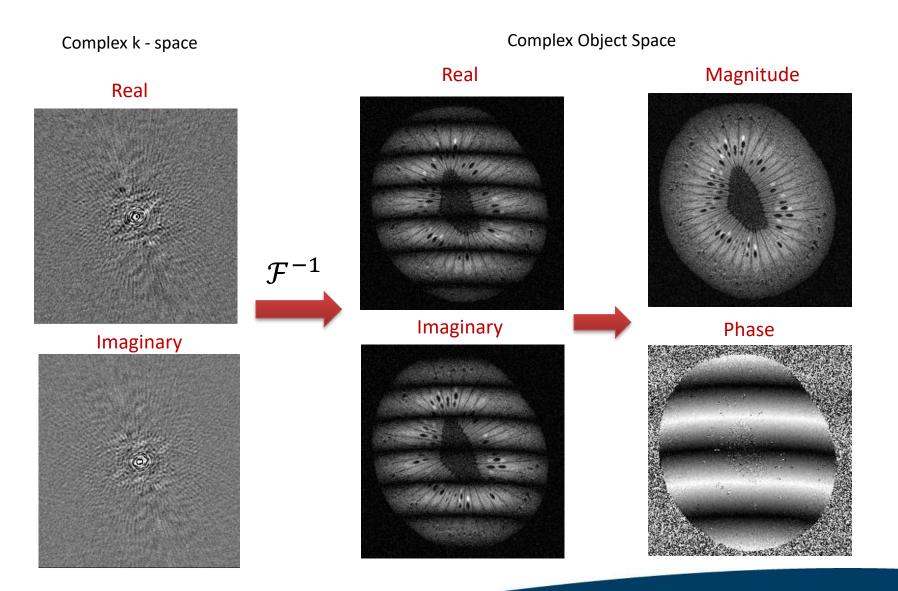
How to measure the level of noise in an image?



Noise Distribution in Magnitude MRI



Noise in MRI – from k-space to Magnitude Images



Noise Distribution in Magnitude MRI

Probability density function

$$p(m|a,\sigma_g) = \frac{m}{\sigma_g^2} e^{-\frac{m^2 + a^2}{2\sigma_g^2}} I_0\left(\frac{ma}{\sigma_g^2}\right) \varepsilon(m) \quad \text{(1)} \quad \text{Rice pdf}$$

at low SNR when a=0



$$p(m, \sigma_g) = \frac{m}{\sigma_g^2} e^{-\frac{m^2}{2\sigma_g^2}} \varepsilon(m)$$
 (2)

Rayleigh

at high SNR

$$p(m|a,\sigma_g) = \frac{1}{\sigma_g \sqrt{2\pi}} e^{-\frac{(m-a)^2}{2\sigma_g^2}}$$
(3)

Gaussian

Experiments and results

Image with Rayleigh distributed background region

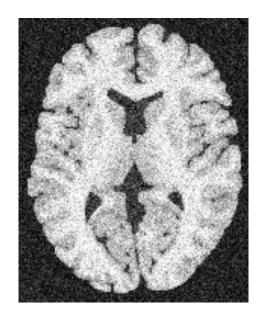
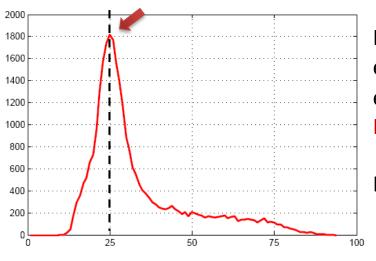
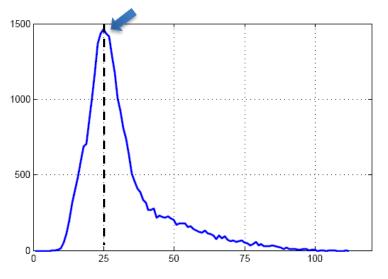


Image corrupted with noise with σ_q =25



Distribution of local estimates of σ_g estimated using local ML method.

Local window size 5 x 5



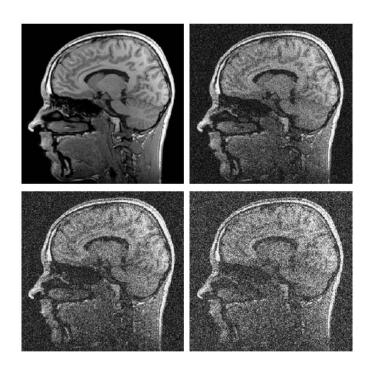
Distribution of local estimates of σ_g estimated based on the local skewness.

Local window size 5 x 5

Image Denoising

Why denoising?

- High noise levels reduce the visibility of small details and lowcontrast changes.
- Generally improves the SNR



Denoising through local averaging

$$I_n(x, y) = I(x, y) + \eta$$

$$I(x, y) = \frac{1}{|N(x, y)|} \sum_{j \in N(x, y)} I(x_j, y_j)$$



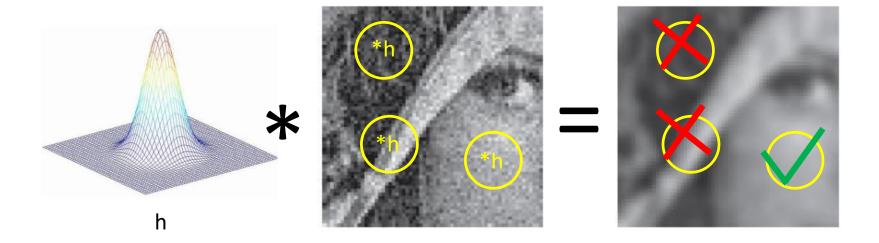
Simple averaging filter – will cause blurring of edges and textures in the image



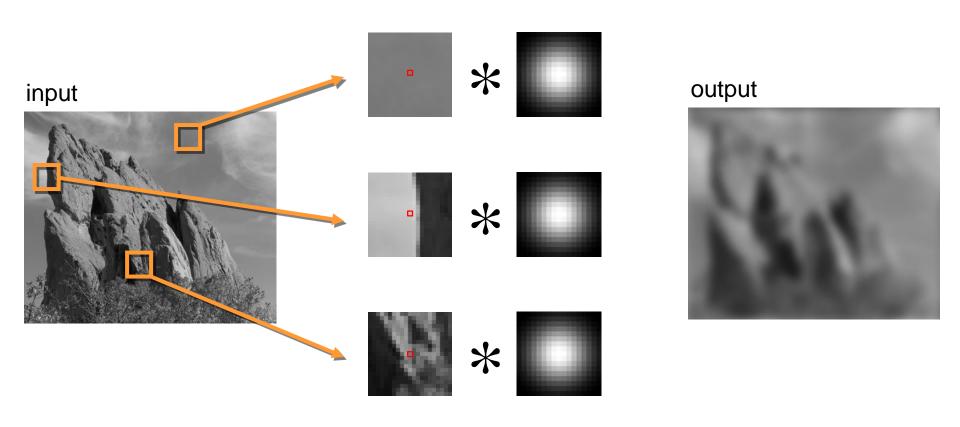




Gaussian Smoothing



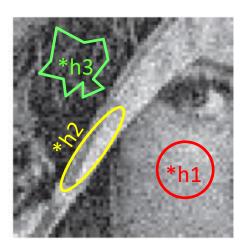
Gaussian Smoothing



Same Gaussian kernel everywhere Averages across edges ⇒ blur

Local adaptive smoothing

- Non uniform smoothing
 Depending on image content:
 - Smooth where possible
 - Preserve fine details

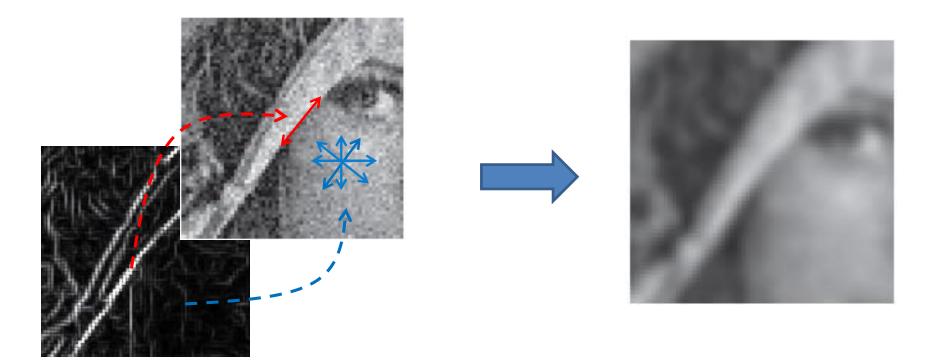


How?

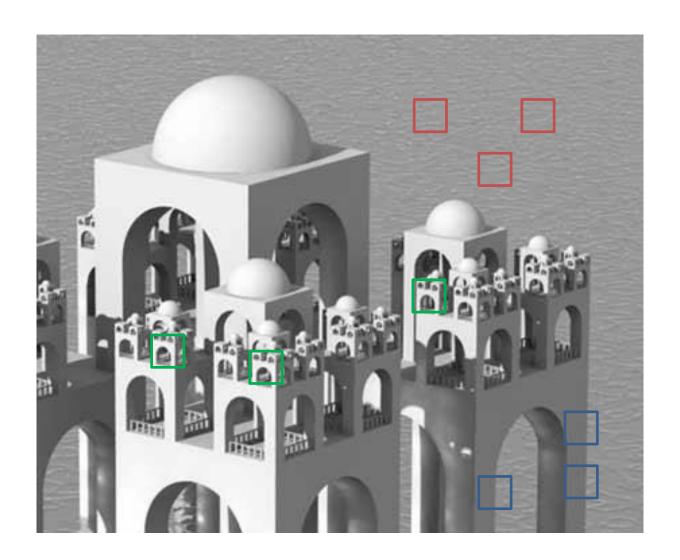
Anisotropic Filtering

gradient

- Edges ⇒ smooth only along edges
- "Smooth" regions ⇒ smooth isotropically



Redundancy in natural images

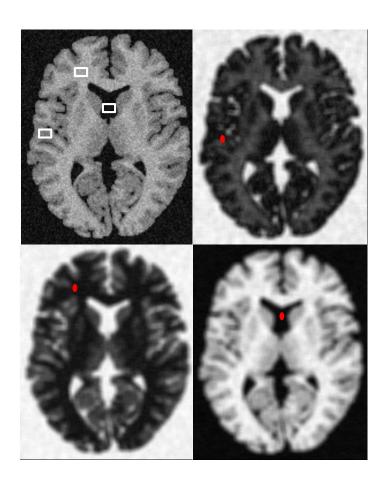


Patch Similarity – Euclidean Distance

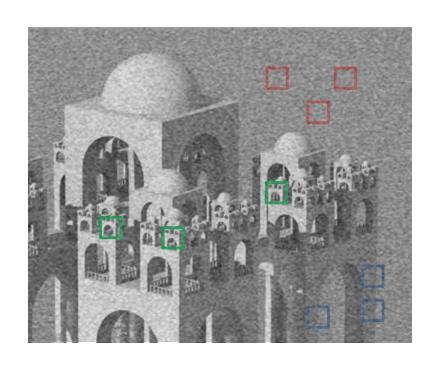
Selecting samples m₁, m₂, m₃ ... m_n

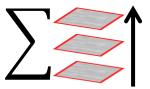
$$d_{i,j} = \|N_i - N_j\|$$

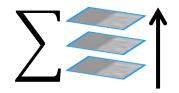
 NL pixels – selected based on the intensity similarity of pixel neighborhoods.



Single Image "time-like" denoising







Unfortunately, patches are not exactly the same ⇒ simple averaging just won't work

Non Local Means – Basic Principle

- Non-local means compares entire patches (not individual pixel intensity values) to compute weights for denoising pixel intensities.
- Comparison of entire patches is more robust, i.e. if two patches are similar in a noisy image, they will be similar in the underlying clean image with very high probability.

Non Local Means (NLM)

Baudes *et al.* (2005) Use a weighted average based on similarity

$$NL(v)(i) = \sum_{j \in I} w(i, j)v(j),$$

$$w(i,j) = \frac{1}{Z(i)} e^{-\frac{\|v(\mathcal{N}_i) - v(\mathcal{N}_j)\|_{2,a}^2}{h^2}},$$

$$Z(i) = \sum_{j} e^{-\frac{\|v(\mathcal{N}_i) - v(\mathcal{N}_j)\|_{2,a}^2}{h^2}}$$

