

Lecture 4: Nonlinear Filters



Image Noise

- Physical effects (dust, raindrops)
- Electrical noise
 - Caused by **fluctuations** in electric circuit.
 - **Thermal** noise
 - **Impulse** noise
- **Quantization** noise
- Poor illumination



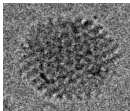
Source: <http://photographylife.com/what-is-iso-in-photography>

Applications

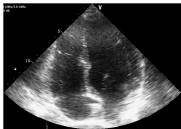
Astronomy



Microscopy



Medical Imaging



Computer Animation



Noise vs. Blur

- Noise adds static by randomly corrupting individual pixels.
- Blur smooths the edges by averaging over multiple pixels.

$$g = f + RANDOM$$



$$g = f * w$$



- Mathematically speaking, noise is generally an additive process, while blur is generally multiplicative.

Artificial Noise

- We can add artificial noise to images using the Matlab `imnoise` command.
- We have to specify what **type** of noise we want and **how much**.
- The **most common** type of noise is **Gaussian** (white noise).

```
A = imread('cameraman.tif');  
B = imnoise(A, 'gaussian', 0, 0.3);
```

Distribution of noise

Mean

Variance



Additive Noise

Gaussian Noise (**MRI**)

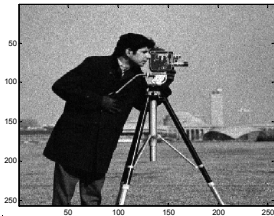
$$f = g + N(0, \sigma)$$



```
A=imread('cameraman.tif');  
B=imnoise(uint8(A), 'gaussian', 0, 0.02);  
imagesc(B);
```

Poisson Noise (**PET**)

$$f = \text{Poisson}(g)$$



```
A=imread('cameraman.tif');  
B=imnoise(uint8(A), 'poisson');  
imagesc(B);
```

Non-additive Noise

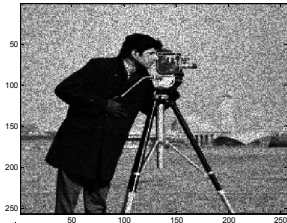
Salt & Pepper Noise

(Reset random pixels)



```
A=imread('cameraman.tif');  
B=imnoise(uint8(A),'salt & pepper',0.1);  
imagesc(B);
```

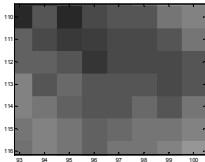
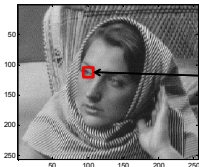
Speckle Noise (Ultrasound)



```
A=imread('cameraman.tif');  
B=imnoise(uint8(A),'speckle',0.05);  
imagesc(B);
```

Filters

- A filter is a process that **removes** or **enhances** some **feature** of an image.
- Commonly, the word “filter” describes an **operation** on the **neighborhood** of an image.



Linear vs. Nonlinear

- A linear filter can be expressed as a convolution (weighted average over a neighborhood):

$$g = f * w$$

- Any filter that cannot be written as a convolution is called a nonlinear filter.
- Linear filters can be computed very quickly using Fourier Transforms by The Convolution Theorem.
- Nonlinear filters will generally be much slower.

Median Filter

- The median filter computes the median of the neighborhood centered over the pixel.
- Compared to the mean filter, the median filter preserves edges better.

255	255	0
255	100	0
255	255	0

Mean

$$= \frac{255+255+0+255+100+0+255+255+0}{9} \\ = 152.8$$

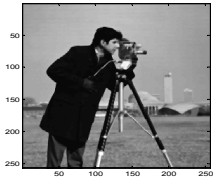
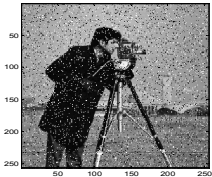
Median

$$= 0, 0, 0, 100, 255, 255, 255, 255, 255 \\ = 255$$

Median Filter

- The median filter and its variants are particularly well-suited for salt & pepper noise.
- The Matlab command `nlfilter` calculates nonlinear functions on neighborhoods. It is much slower than `imfilter`.

```
B = nlfilter(A, [3,3], @(x) median(x(:)) );
```



- Can also use the command `medfilt2`.

Visualizing Noise

- A common way to test denoising algorithms is to subtract the denoised image from the noisy image.
- This subtracted image is called the residual.
- If the algorithm did a good job denoising, then we should only see random dots and no structure.

Noisy

Denoised by Median Filter

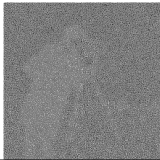
Residual



-



=



Quantifying Noise

- There are two major statistics used to determine the amount of noise in an image.
- Both statistics compare the noisy image f to an ideal noise-free image f_{ideal} .
- Both statistics are easier to write using the Frobenius norm:

$$\|A\|_F = \sqrt{\sum_i \sum_j f(i,j)^2}$$

1. Root Mean Square Error (RMSE)

$$RMSE = \frac{1}{N} \|f - f_{ideal}\|_F \quad \text{where } N = \text{\#pixels in image}$$

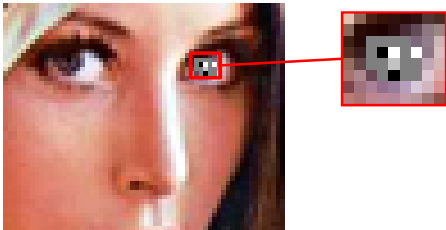
2. Signal-to-Noise Ratio (SNR)

$$SNR = 20 \log \frac{\|f_{ideal}\|}{\|f - f_{ideal}\|}$$

Larger SNR = Less Noise

Beyond Local

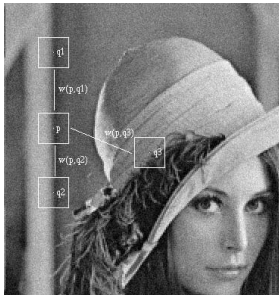
- Suppose Lena's left eye was obscured by noise.
- A neighborhood around the eye will not tell us the eye's true color.



- *Solution:* Look at the other eye!

Image Similarity

- Natural images tend to have similar neighborhoods.



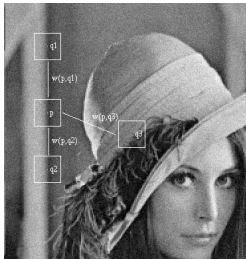
Nonlocal-Means

- *(Buades-Coll-Morel, 2005)* proposed a denoising method that averages over all pixels in the image that have a similar neighborhood.
- They call their method *Nonlocal (NL) Means*.

$$NL(x) = \frac{\sum_q w(x, q) f(q)}{\sum_q w(x, q)}$$

where w is a weight that judges the similarity between the neighborhoods.

Similar $\rightarrow w=1$, Different $\rightarrow w=0$



Nonlocal-Means

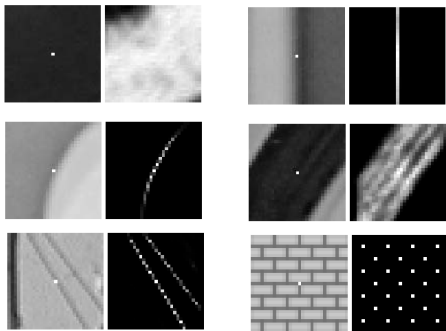
- Let $N(x)$ denote the neighborhood around pixel x .
- We want $w(x,q)=1$ if the neighborhoods $N(x)$ and $N(q)$ are similar.
- And we want $w(x,q)=0$ if the neighborhoods $N(x)$ and $N(q)$ are completely different.

$$w(x, q) = \exp\left(-\frac{\|N(x) - N(q)\|_F^2}{h^2}\right)$$

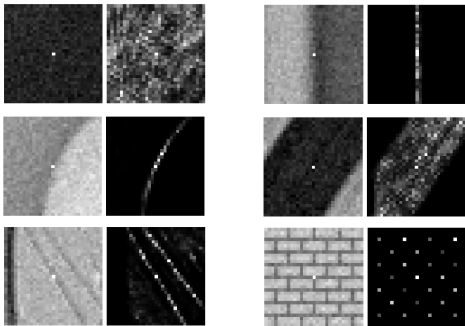


- The parameter h controls the similarity.
- Generally choose square neighborhood, e.g. 5x5.
- Can add Gaussian weights to norm to emphasize center.

Weights from Clean Images

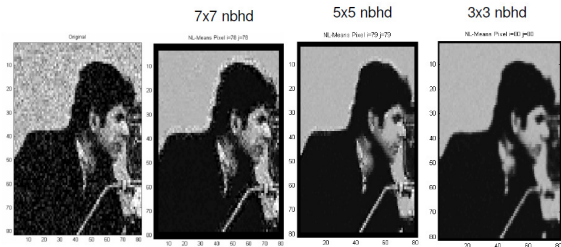


Weights from Noisy Images



NL-Means

- Smaller neighborhoods remove noise better, but may over-smooth the image.



NL-Means

- Many consider NL-Means and its variants to be the best denoising method for natural images.



Noisy



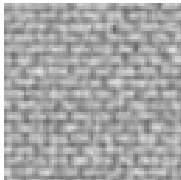
NL-Means

NL-Means

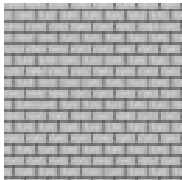
- NL-Means performs **best** on images with **repeating patterns**.



Noisy



Gauss Filtered



NL-Means



Implementation

- NL-Means is **painfully slow**. There are 2 tricks to speed it up:
 1. *Search Somewhat Local* -- **Restrict the neighborhood comparisons to a window around the current pixel**, rather than the whole image.
 2. *Save Pixel Comparisons* -- **Store all pixel-to-pixel comparisons in an efficient data structure** to avoid making unnecessary passes through the image

Color Images

- We can extend NL-Means to color images by comparing 3D neighborhood cubes.
- NL-Means actually achieves better results on color images than on grayscale. We can shrink the neighborhood size because of the extra information.



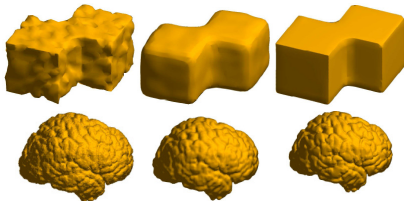
Noisy



NL-Means

3D NL-Means

- NL-Means extends to 3D for surface denoising (*Dong-Ye-Osher-Dinov, 2007*).
- NL-Means recovers corners and flattens smooth regions better than many denoising methods.



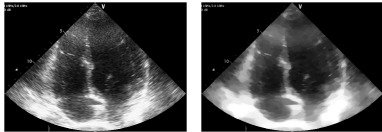
Noisy

Mean Curvature Smoothing

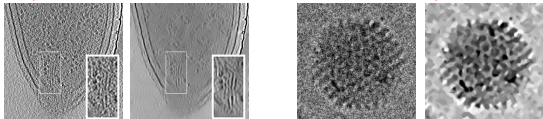
NL-Means

Ultrasound

- Nonlocal methods are good at removing the speckle noise in Ultrasound images (*Sawatzky, 2011*).

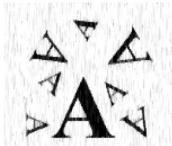


- NL-Means is particularly good for microscopy (*Darbon-Cunha-Chan-Osher-Jensen, 2008*).



More Neighborhoods

- Standard NL-Means only compares neighborhoods under translation.
- We could also consider neighborhoods under scaling and rotation transformations (*Lou-Favaro-Soatto-Bertozzi, 2008*).
- But this will slow down the computation.



Noisy



NL-Means



Extended NL-Means



Use with Caution

- Nonlocal methods are powerful denoisers, but they have to be used carefully.
- Exploiting image similarity could erase small-scale features that only occur in one part of the image.
- But these small features are often the most important part in some applications. For example, in medical imaging we look for tumors, hemorrhages, muscle tears, bone fractures, etc.