### Image Denoising

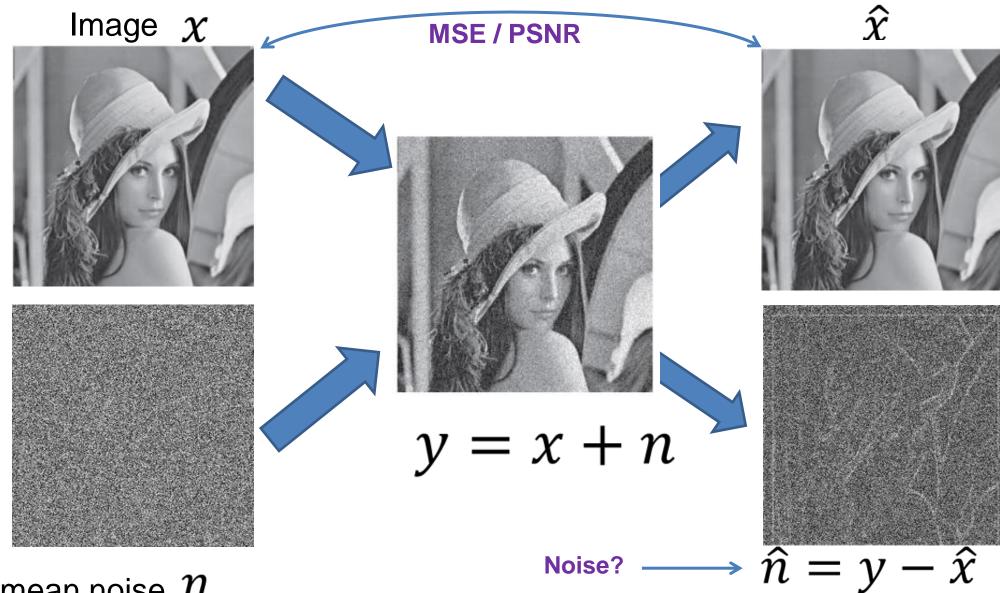
Beyond Blurring and Median Filtering







#### Problem Definition: Noise Cleaning



Zero mean noise  $\,n\,$ 

#### Quality Measures for Restoration

- Given original image I(x,y), restored to  $\hat{I}(x,y)$
- In real life I(x,y) is unknown, but known in testing.
- Assumption: Additive White Gaussian Noise (AWGN)
- Mean Square Error

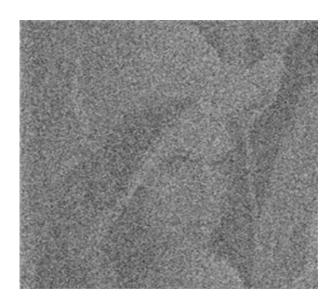
$$MSE = \frac{1}{N^2} \sum_{(x,y)} ||I(x,y) - \hat{I}(x,y)||^2$$

Peak Signal to Noise Ratio

$$PSNR = 20log_{10} \frac{255}{\sqrt{MSE}}$$

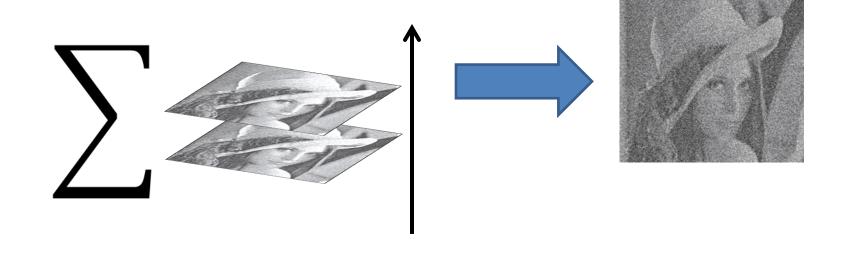
# Patch Methods: Non-Local Means Motivation - Temporal perspective

- Assume a static scene, giving a constant signal x(t)
- Multiple images y(t) are captured at different times y(t)=x(t)+n(t)
- Noise n(t) varies over time with zero mean



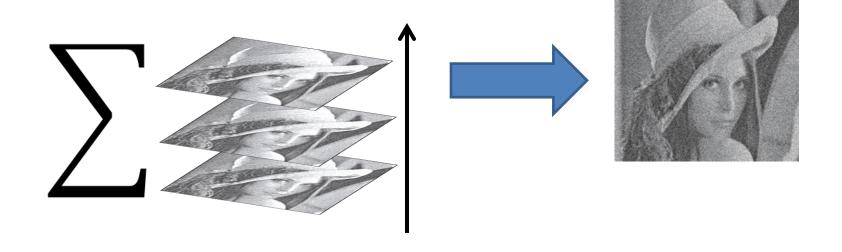
#### Temporal Denoising

Stationary Images - Average multiple images over time



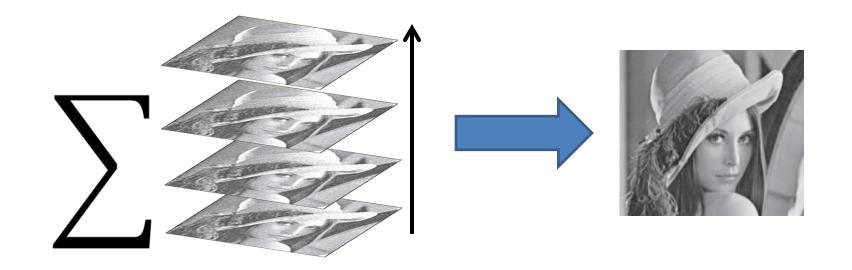
#### Temporal Denoising

Stationary Images - Average multiple images over time



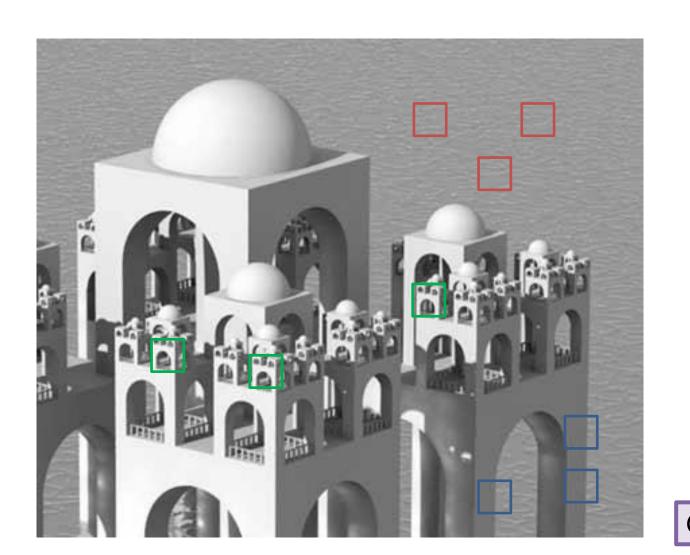
#### Temporal Denoising

Stationary Images - Average multiple images over time



It works because (i) the noise is zero mean (ii) When variance of each variable  $\{X_i\}_{i=1}^n$  is  $\sigma^2$ , the variance of their average is  $\sigma^2/n$   $Var(\bar{X}) = Var\left(\frac{1}{n}\sum_{i=1}^n X_i\right) = \frac{\sigma^2}{n}$ 

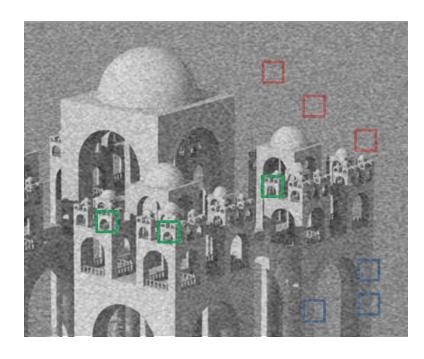
## Redundancy in Natural Images Many image patches are similar

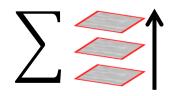


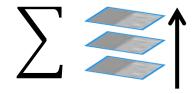
Glasner & Irani (2009)

### Single Image Denoising Find and Average Similar Patches









Use weights as patches are not identical

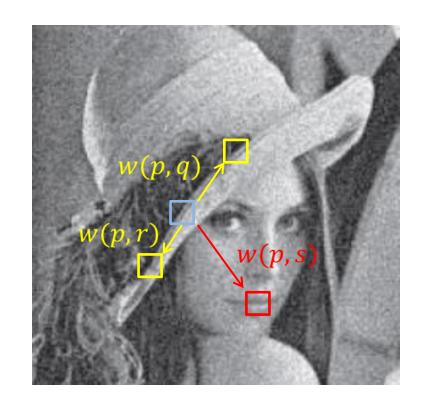
#### Non-Local Means (NLM)

- Given one pixel, compute the similarity of a patch around it to patches around **all other** pixels.
- Compute a weighted average of pixels based on patch similarity.
- Replace pixel value by this average

$$\hat{x}(m,n) = \frac{1}{C(m,n)} \sum_{i,j} y(i,j) e^{-(SSD(N(m,n)-N(i,j)))}$$

$$W_{mnij}$$

y is input image,  $\hat{x}$  is output image, N(i,j) is a neighborhood around pixel N(i,j)



#### Non-Local Means Equation

$$\hat{x}(m,n) = \frac{1}{C(m,n)} \sum_{i,j} y(i,j) w(m,n,i,j)$$

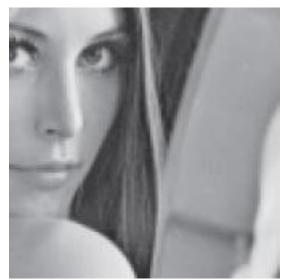
$$w(m,n,i,j) = e^{-\frac{SSD(N(m,n)-N(i,j))}{2\sigma^2}}$$

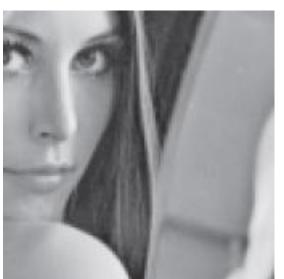
- The pixel value at location (m,n) is the average of all other pixels (i,j) in the image, weighted by w(m,n,i,j)
- The weights are computed from Sum of Square Differences(SSD) between neighborhoods of (m,n) and of (i,j), N(m,n) etc.
  - SSD can have equal weights for all pixels in N(m,n)
  - Or: Gaussian weights, where center pixel has higher weight
  - Or: neighborhoods can be normalized by mean and variance

#### Patch Based Noise Cleaning Methods

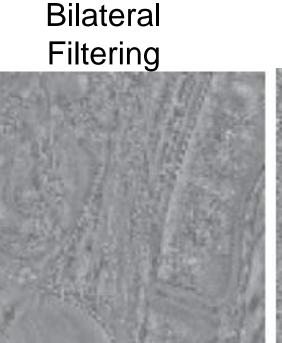
- NLM is one of a family of patch-based noise cleaning methods
- Variations include: Other similarity measures between patches, replacing the SSD
- Define search areas for patches, E.g. search also in other frames (Google's Night-Sight)
- Methods to replace averaging, E.g. Multi-frame superresolution (Google's Night-Sight)

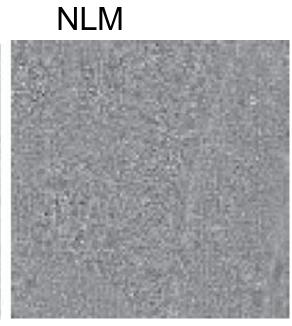
## Comparison Difference Between Noisy and Denoised







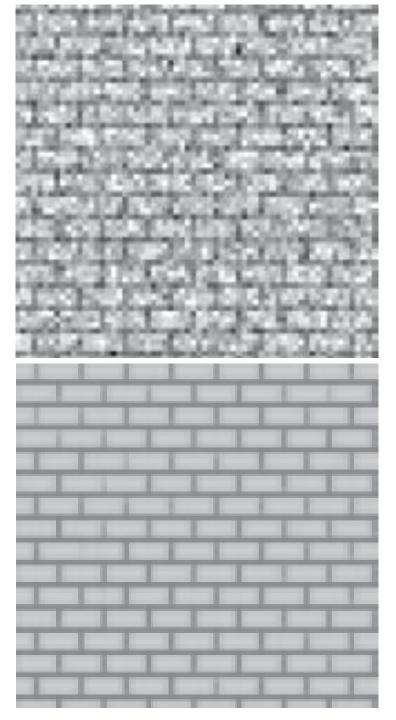








#### NLM

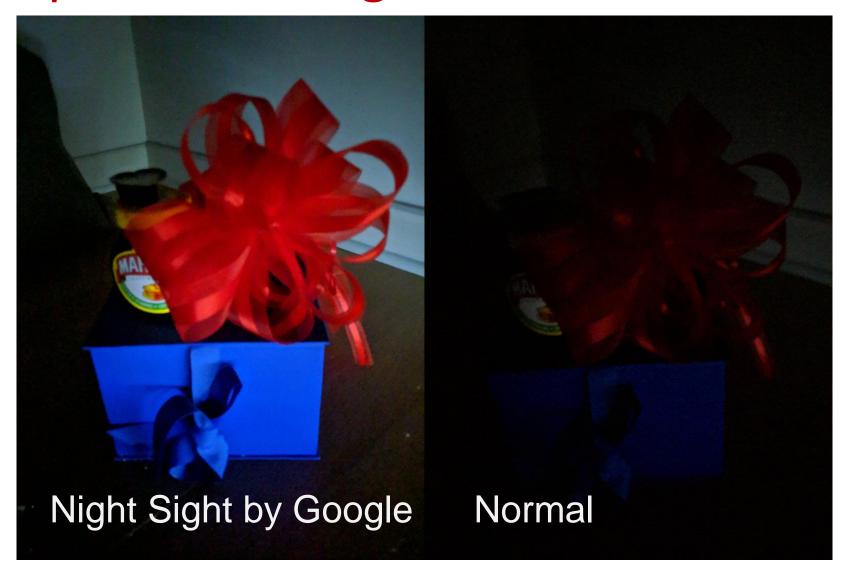


#### NLM

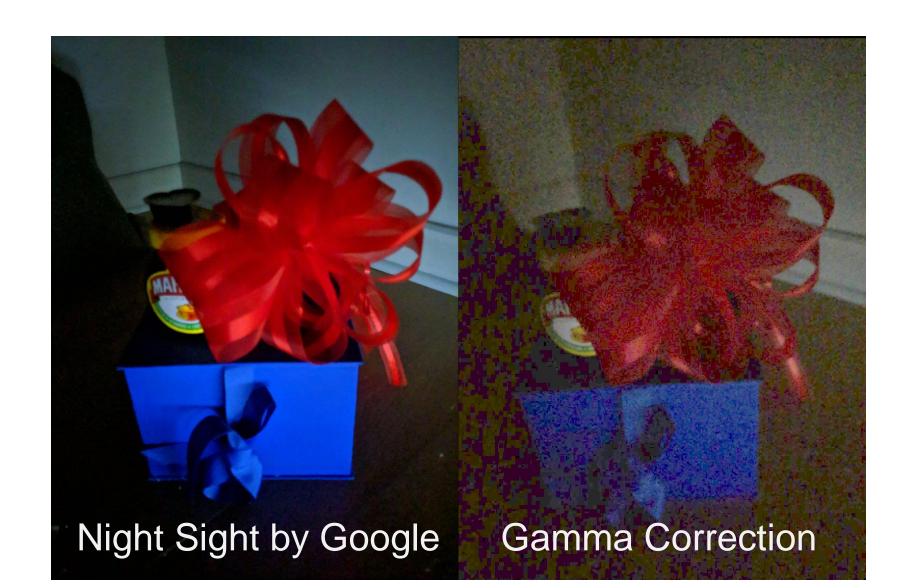




## "GOOGLE Gives the Pixel Camera Superhuman Night Vision" 14/11/18



#### Increasing Lightness by Gamma Correction



### Google's "Night Sight" Process by Yael Pritch

- Low light traditional approach: Use very long exposure
- Take multiple pictures, 6 Sec. (6-15 pictures)
- Exposure time determined from motion (gyro & OF, Next...)
- Before averaging: Perform alignment between pictures to compensate for motion (Next in course)
- Combine overlapping patches using Multi-frame Superresolution approach (Read...)
- Color Correction ("White Balance" Read...)
- Problem: How to focus?

