Advanced Image Denoising Methods: TV, NLM, and BM3D

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Outline

• TV [Rudin, et al. 1992]:

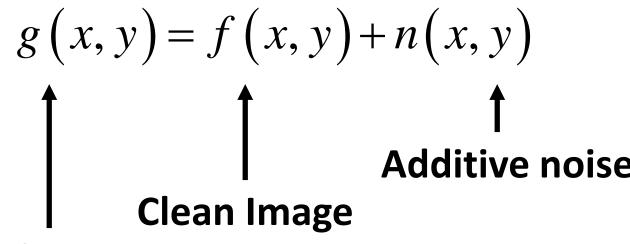
Total Variation minimization

• **NLM** [Buades, et al. 2005]:

Non-Local Means

BM3D [Dabov, et al. 2006/2007]:

Block-Matching and 3D filter



Observed image

$$\hat{f} = g \otimes h$$
 —— Smooth kernel (local method)

Global method:

$$TV(\hat{f}) = \sum_{x,y} \sqrt{(\hat{f}(x+1,y) - \hat{f}(x,y))^2 + (\hat{f}(x,y+1) - \hat{f}(x,y))^2}$$





$$\arg\min_{\hat{f}} \frac{1}{2} \left\| g - \hat{f} \right\|_{2}^{2} + \lambda \cdot TV(\hat{f})$$

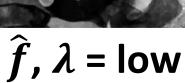
f =



Fidelity term









$$\hat{f}$$
, λ = high

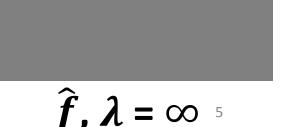




Figure 1. An image.

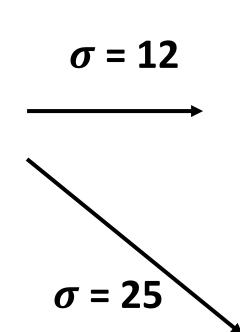




Figure 2. The image of Fig. 1 and its reconstruction ($\sigma = 12$).



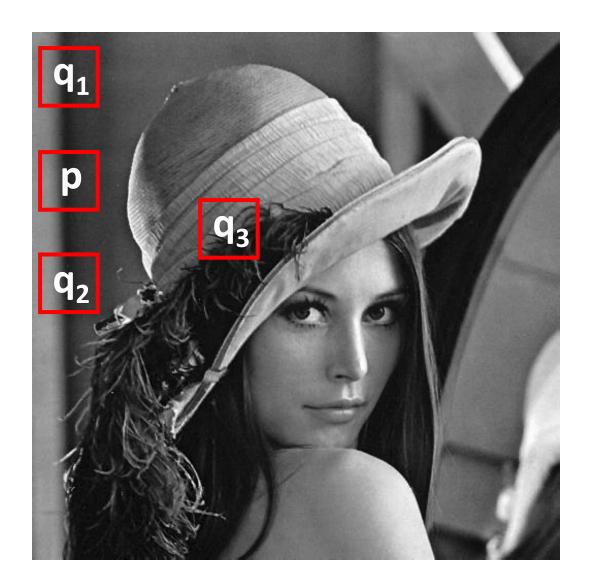
- Straight edges are maintained.
- Details and texture can be over smoothed if λ is too large.





Figure 3. Same as Fig. 2 with now $\sigma = 25$.

Non-Local Means (NLM)



$$NLM(x, y) = \sum_{i,j} w_{xy}(i, j)g(i, j)$$

$$0 \le w_{xy}(i,j) \le 1, \sum_{i,j} w_{xy}(i,j) = 1$$

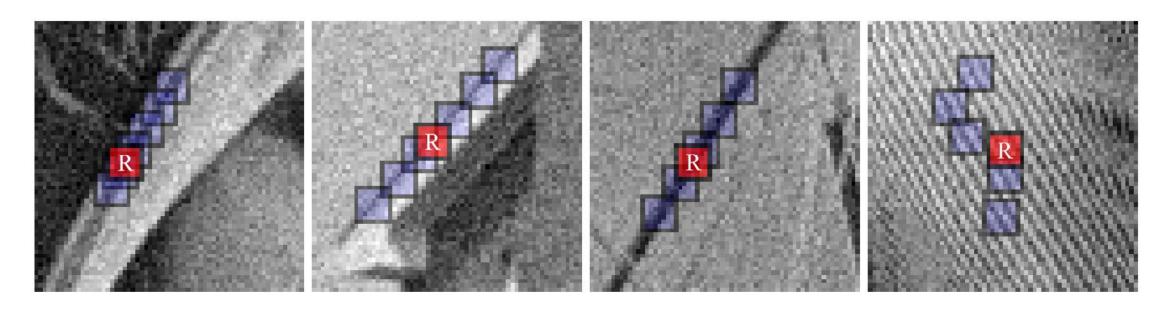
$$w_{xy}(i,j) = \frac{\exp\left(-\frac{\left\|g(N_{xy}) - g(N_{ij})\right\|_{2}^{2}}{\sigma^{2}}\right)}{\sum_{ij} \exp\left(-\frac{\left\|g(N_{xy}) - g(N_{ij})\right\|_{2}^{2}}{\sigma^{2}}\right)}$$

Non-Local Means (NLM)

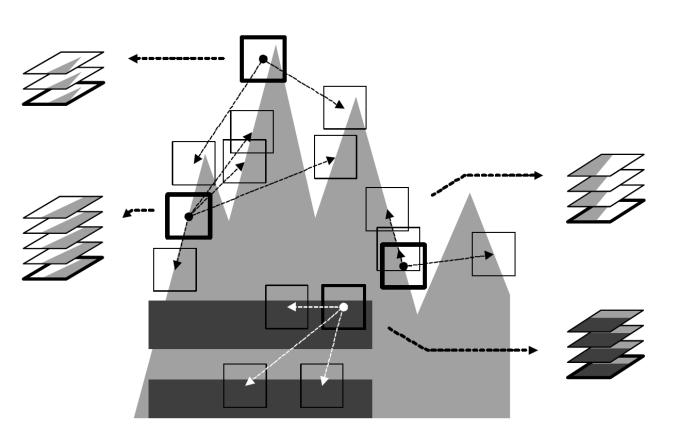


Preserve straight edges, as well as details and texture.

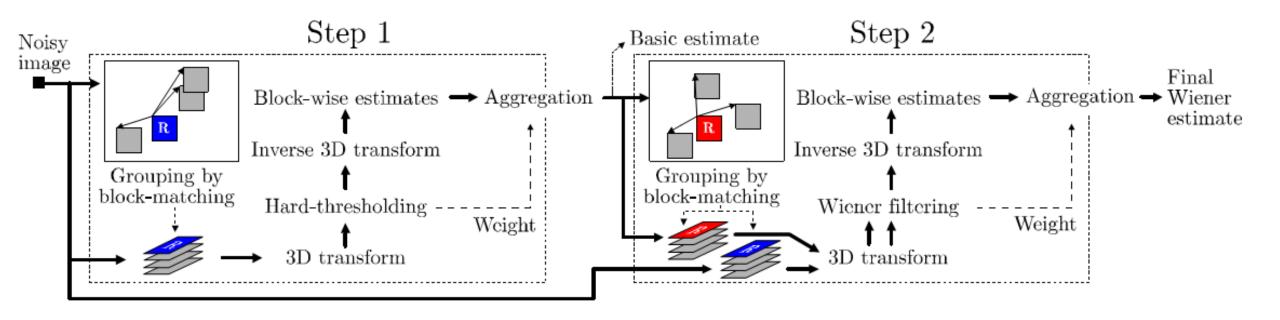
Block matching + 3D transform



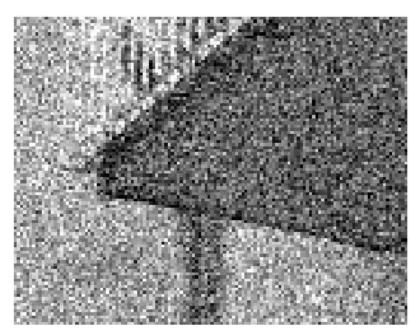
Block matching + 3D transform



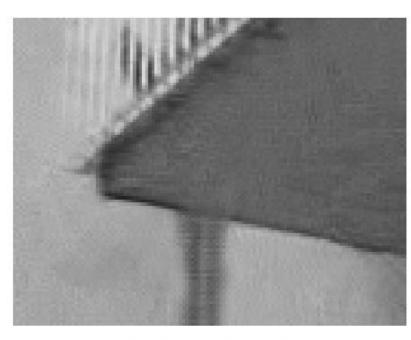
- Element-wise averaging
 - Identical blocks
 - Multiple blocks
- 3D transform (e.g., DWT, DFT, DCT)



- Using the basic estimate instead of the noisy image allows to improve the grouping by block-matching.
- Using the basic estimate as the pilot signal for the empirical Wiener filtering is much more effective and accurate than the simple hard-thresholding.



Noisy image $\sigma = 40$

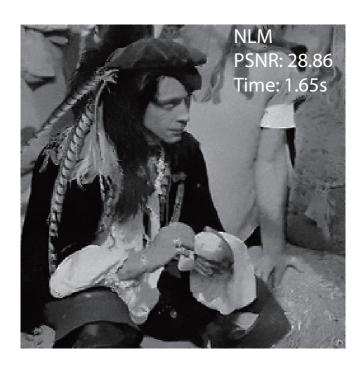


Basic estimate



Final estimate







Thank How