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O L L S C O I L L U I M N I G H

Perona-Malik Anisotropic Filtering

Machine Vision – RE4107
Dr. Colin Flanagan

Group Members:

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Introduction

Perona-Malik diffusion or anisotropic diffusion is a computer vision filtering technique that aims at reducing image noise without reducing the image's quality in the process. This diffusion technique typically resembles the process that creates a scale space, where an image generates a parameterized family of successively more and more blurred images based on a diffusion process. Each of the resulting images in this filter is given as a convolution between the image and a 2D isotropic Gaussian filter. This diffusion is a linear and space invariant transformation of the original image. Anisotropic diffusion is a much generalized version of the diffusion process. It produces a family of parameterized images where each resulting image is a combination between a filter that depends on the content of the original image and the original image itself. This means that anisotropic diffusion is a linear and space invariant transformation of the original image.

Algorithm Properties

Let $\Omega \subset \mathbb{R}^2$ denote a subset of the plane and $I(\cdot, t) \rightarrow \mathbb{R}$ be a family of greyscale images. With this, anisotropic diffusion can be defined as:

$$\frac{\partial I}{\partial t} = \text{div} (c(x, y, t) \nabla I) = \nabla c \cdot \nabla I + c(x, y, t) \Delta I$$

Where, Δ denotes Laplacian, ∇ denotes Gradient, $\text{div}(\dots)$ is the divergence operator and $c(x, y, t)$ is the diffusion coefficient. $c(x, y, t)$ also controls the rate of the diffusion. When Perona and Malik pioneered the idea they proposed two functions for the diffusion coefficient:

$$c(\|\nabla I\|) = e^{-(\|\nabla I\|/K)^2}$$

and

$$c(\|\nabla I\|) = \frac{1}{1 + \left(\frac{\|\nabla I\|}{K}\right)^2}$$

The constant K controls the sensitivity to edges and is usually chosen experimentally or as a function of the noise in the image.

Comparison of “Stopping” Functions

To implement an alternate stopping function, the following code should be implemented in the python program:

```
def f(lam,b):  
    func = 1/(1 + ((lam/b)**2))  
    return func
```

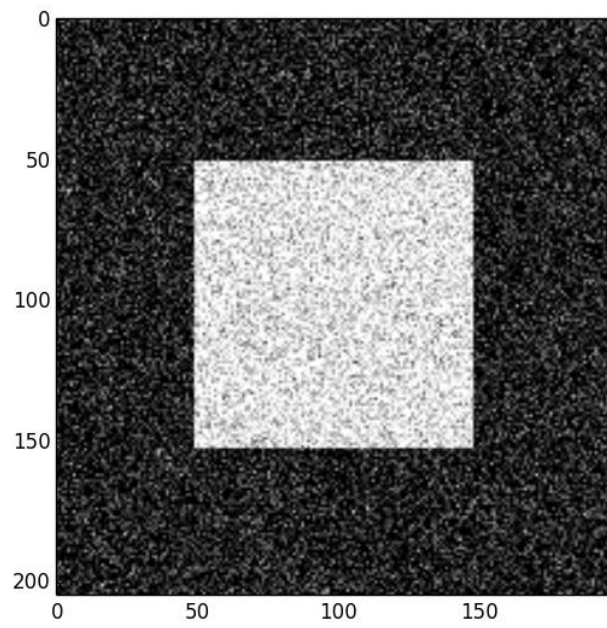
This code is commented into our program.

Procedure

This section contains the input and output from our python code. By viewing these images it is easy to see Perona-Malik anisotropic filtering in action. The images were taken from the sample images provided in the Sulis folder.

Image 1: noisy_rect.png

Before



After

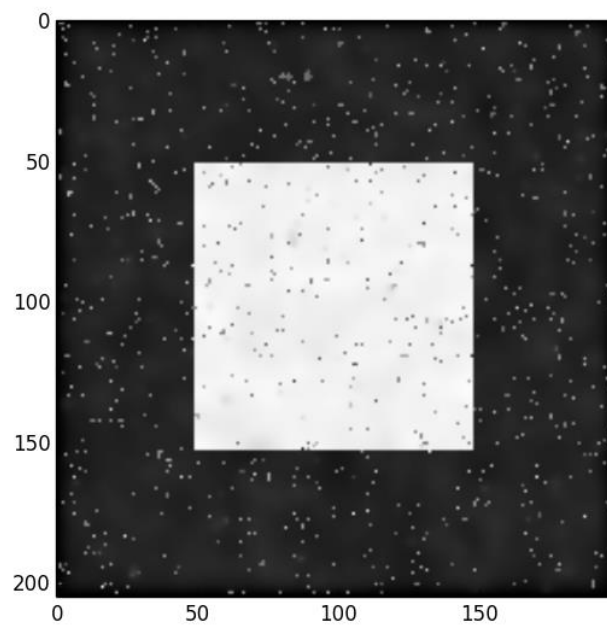
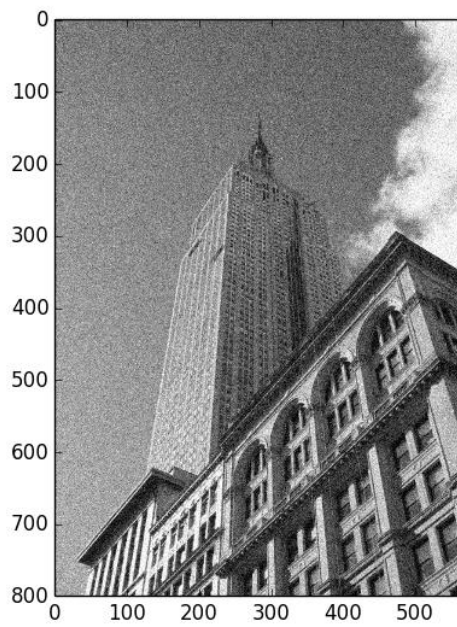
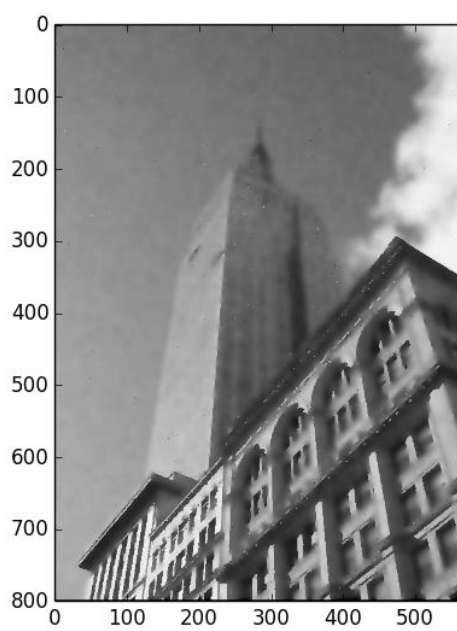


Image 2: noisy_empire.png

Before



After



Conclusions

It is obvious that the above images were successful in efficiently implementing Perona-Malik anisotropic diffusion. The above images clearly show a significant noise reduction difference between the before and after images. Noise reduction was successfully achieved but it is also important to note the edges and lines maintained their sharpness.