

## An Efficient Operator-Splitting Method for Noise Removal in Images

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Received 5 October 2005; Accepted (in revised version) 5 March 2006

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**Abstract.** In this work, noise removal in digital images is investigated. The importance of this problem lies in the fact that removal of noise is a necessary pre-processing step for other image processing tasks such as edge detection, image segmentation, image compression, classification problems, image registration etc. A number of different approaches have been proposed in the literature. In this work, a non-linear PDE-based algorithm is developed based on the ideas proposed by Lysaker, Osher and Tai [*IEEE Trans. Image Process.*, 13 (2004), 1345-1357]. This algorithm consists of two steps: flow field smoothing of the normal vectors, followed by image reconstruction. We propose a finite-difference based additive operator-splitting method that allows for much larger time-steps. This results in an efficient method for noise-removal that is shown to have good visual results. The energy is studied as an objective measure of the algorithm performance.

**Key words:** Noise removal; nonlinear PDEs; additive operator splitting (AOS).

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## 1 Introduction

In this work, removal of additive, zero-mean noise in digital images is investigated. We use the ideas proposed in [1], based on a TV-norm approach. This results in two nonlinear partial differential equations. The first of these equations is the smoothing of the flow-field (normal field) of the original image. The second equation reconstructs a noise-reduced image from the smoothed flow-field. This results in an efficient method for noise-removal that has good results.

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The contributions of this paper are as follows: a more efficient scheme for flow-field smoothing is developed; an additive operator splitting (AOS) method ([2, 15, 16]) is employed to further improve the efficiency of the flow-field smoothing; a comparison of the AOS method and explicit methods are done in terms of numerical performance; and lastly, the use of the energy is suggested as an objective measure of the performance of noise-reduction algorithms based on energy minimization.

Let  $d$  be a digital image defined on a two-dimensional region  $\Omega$ . Let  $(x, y)$  denote the position of a single pixel.  $d(x, y)$  is the grey-level value associated with the pixel  $(x, y)$ . The noise model is assumed to be zero-mean and additive, denoted as  $\eta(x, y)$ . The observed image values  $d_0(x, y)$  are

$$d_0(x, y) = d(x, y) + \eta(x, y). \quad (1.1)$$

The problem is to recover the (unknown) true image  $d(x, y)$  from the given observations  $d_0(x, y)$ .

The visually annoying parts of the noise usually belong in the higher frequency regions of the spectrum. A lot of noise can be effectively filtered by a lowpass filter. However, this will remove the true high-frequency components of the image such as edges and texture. Isotropic filters suffer from this problem. The challenge is to retain as much of the true high-frequency information as possible while reducing the perceived noise levels in the image.

The Total Variation (TV) norm based filters proposed in [5], have been shown to be quite effective in removing noise without causing excessive smoothing of the edges. The original formulation of this filter is to obtain  $d(x, y)$  as a solution of the constrained optimization problem

$$\inf_d \int_{\Omega} |\nabla d| \, dx \quad \text{subject to} \quad \int_{\Omega} |d - d_0|^2 \, dx = \sigma^2, \quad (1.2)$$

where  $\sigma^2$  represents the noise level. The resulting Euler-Lagrange PDE to be solved in this case is

$$-\nabla \cdot \left( \frac{\nabla d}{|\nabla d|} \right) + \mu(d - d_0) = 0. \quad (1.3)$$

However, it is well known that the TV norm filter has the disadvantage of a *stair-case effect*: smooth functions get transformed into piecewise constant functions. This lends an undesirable blocky effect to the smoothed image. In [1], it is proposed to modify the equation (1.2). Instead of minimizing the TV norm of  $d$ , it is proposed in [1] to minimize the TV norm of  $\nabla d/|\nabla d|$ , giving the following equation

$$\inf_d \int_{\Omega} \left| \nabla \frac{\nabla d}{|\nabla d|} \right| \, dx \quad \text{subject to} \quad \int_{\Omega} |d - d_0|^2 \, dx = \sigma^2, \quad (1.4)$$

$\vec{n} = \nabla d/|\nabla d|$  is called the normal field of the image. The fourth-order Euler-Lagrange equation that results from directly minimizing this functional is difficult to solve numerically in a stable manner. Therefore, in [1], a re-formulation of this equation is done. The