

## A NEW MEDIAN FORMULA WITH APPLICATIONS TO PDE BASED DENOISING\*

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**Abstract.** We develop a simple algorithm for finding the minimizer of the function  $E(x) = \sum_{i=1}^n w_i |x - a_i| + F(x)$ , when the  $w_i$  are nonnegative and  $F$  is strictly convex. If  $F$  is also differentiable and  $F'$  is bijective, we obtain an explicit formula in terms of a median. This enables us to obtain approximate solutions to certain important variational problems arising in image denoising. We also present a generalization with  $E(x) = J(x) + F(x)$  for  $J(x)$  a convex piecewise differentiable function with a finite number of nondifferentiable points.

**Key words.** Convex optimization,  $\ell_1$  minimization, TV denoising, Bregman iterative method.

**AMS subject classifications.** 46N10, 94A08.

### 1. Introduction

Given  $a_1, \dots, a_n \in \mathbb{R}$ , it is well-known that

$$\min_{x \in \mathbb{R}} \sum_{i=1}^n |x - a_i|^\alpha = \begin{cases} \text{mean}(a_i) & \text{if } \alpha = 2 \\ \text{median}(a_i) & \text{if } \alpha = 1 \\ \text{mode}(a_i) & \text{if } \alpha = 0. \end{cases}$$

More generally, the very early work of Barral [14] investigated

$$\min_{x \in \mathbb{R}} \sum_{i=1}^n w_i |x - a_i|^\alpha, \quad w_i \geq 0, \alpha = 0, 1, 2, \infty.$$

Also, [15, 16] developed local M-estimator filters based on such minimizers.

This work was inspired by two variational problems arising in image research. One is soft wavelet thresholding [2, 8] or basis pursuit [3] arising in compressed sensing. The other is the Rudin-Osher-Fatemi (ROF) model [1] of TV-based image denoising and generalizations. The first involves reducing  $E(x) = \sum_{i=1}^n w_i |x - a_i| + F(x)$  to its simplest terms: the scalar problem

$$\min_{x \in \mathbb{R}} E(x) = |x| + \lambda(x - f)^2, \quad \lambda > 0. \quad (1.1)$$

The solution to (1.1) is obtained from a simple formula:  $x_{opt} = \text{shrink}(f, \frac{1}{2\lambda})$ , where the shrink operator can be found in figure 1.1

It turns out that

$$\text{shrink}(f, \frac{1}{2\lambda}) = \text{median}\{f - \frac{1}{2\lambda}, 0, f + \frac{1}{2\lambda}\}. \quad (1.2)$$

This seems (surprisingly) to be a new result. We generalize it below.

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