# Image Denoising with a Constrained Discrete Total Variation Scale Space

Igor Ciril, LMCS, Institut Polytechnique des Sciences Avancées (IPSA) Jérôme Darbon, CNRS / CMLA, Ecole Normale Supérieure Cachan

#### Contribution

We consider a combinatorial approach that relies on coupling the TV-flow (which corresponds to the solution of a differential inclusion) that incrementally simplifies the original noisy image with a procedure that intends to recover the contrast.

#### Notations

- Markovian framework:
  - set of pixels :  $\mathcal{V}$

#### A coupled scale-space approach

Approach coupling two procedures

- 1. Procedure of simplification (denoising but loss of contrast) of the observed image:  $t \mapsto \mathbf{u}(t)$  solution of DTV-flow
- 2. Procedure that respects shapes and recovers the contrast:  $t \mapsto \tilde{\mathbf{u}}(t)$  is the image that is the closest to the observed image **f** having the same relative order as u(t). This corresponds to the projection of **f** onto the convex set:

$$\bigcap_{(j)\in\mathcal{W}} \left\{ \mathbf{g}\in\mathbb{R}^N \mid \underbrace{|g_j - g_i| + m_i(\partial R_{i,j}(\mathbf{u}(\mathbf{t})))(g_j - g_i)}_{\text{relative order for }(i,j)} = 0 \right\}$$

- value of image u at site *i*:  $u_i$
- set of interactions:  $\mathcal{W}$
- Discrete Total Variation (DTV)

$$J(\mathbf{u}) = \sum_{(i,j)\in\mathcal{W}} R_{i,j}(\mathbf{u}) = \sum_{(i,j)\in\mathcal{W}} |u_j - u_i|$$

- Sub-differential of *F* at *x*
- $\partial F(\mathbf{x}) = \{ \mathbf{s} | \forall \mathbf{y}, \langle \mathbf{y} \mathbf{x}, \mathbf{s} \rangle + F(\mathbf{x}) \leq F(\mathbf{y}) \}$
- Minimal subgradient of *F* at *x*

 $m(\partial F(x)) =$ projection of 0 onto  $\partial F(x)$ 

#### **Discrete Total Variation Flow**

• DTV-flow (Differential Inclusion)

 $\frac{d\mathbf{u}}{dt}(t) \in -\partial J(\mathbf{u}(t))$  on  $(0, +\infty)$ 

### Results



(a) Original image





(b) Noisy image



## $\mathbf{u}(0) = \mathbf{f}$

- The slow solution of the differential inclusion yields the trajectory of DTV-flow
- Computed exactly using a network-flow approach
- It generates a sequence of images that simplfies more and more the original image as time evolves
- It presents a **loss of contrast**

 $\Rightarrow$  Idea: get back the constrast while preserving the geometric information

#### **Relative Order Preservation**

- We want to keep the relative order of the level lines
- This constraint is maintained through:

(a) Our result





(b) Residual



- constraining relative order between two interacting pixels
- using Bregman distances

 $|u_j - u_i| + m_i(\partial R_{i,j}(\mathbf{v}))(u_j - u_i) = 0$ 

 $\Rightarrow$  Geometric information maintained as a variational form

- Need to select the minimal subgradient:
  - otherwise relative order not necessarily satisfied
  - required for convergence properties of the approach

(a) TV minimizer

(b) Residual

Funding Research of J. Darbon has been supported by US Office of Naval Research ONR N000140710810.