

A Directional Total Variation

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Aims of the Work

The aims of this work are:

- ▶ To define a directional TV – modified TV that is sensitive to directions .
- ▶ To demonstrate the utility of the directional TV.
 - ▶ To develop an algorithm for the implementation of directional TV for image denoising problem.
 - ▶ To demonstrate the performance of the directional TV against regular (isotropic) TV on images with a dominant direction.

Total Variation (TV)

- ▶ TV is a commonly used prior for images
- ▶ TV of a discrete-space image f is defined as,

$$\text{TV}(f) = \sum_{i,j} \|\Delta f(i,j)\|_2$$

or

$$\text{TV}(f) = \sum_{i,j} \sup_{t \in B_2} \langle [\Delta f(i,j), t] \rangle$$

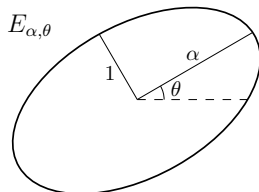
where Δ is defined as,

$$\Delta f(i,j) = \begin{bmatrix} f(i,j) - f(i-1,j) \\ f(i,j) - f(i,j-1) \end{bmatrix}$$

and B_2 is the unit ball of the ℓ_2 norm.

Directional TV

- ▶ TV is isotropic as it is invariant under a rotation in the image.
- ▶ It is possible to obtain directional TV by replacing B_2 with some other set.
- ▶ Replace B_2 with an ellipse, $E_{\alpha,\theta}$, that is characterized with the ratio of major axis to the minor axis (α) and the angle of orientation (θ).

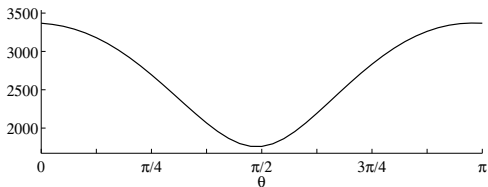
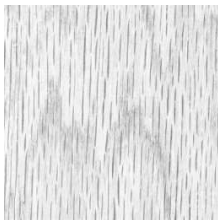


- ▶ The resulting norm is more sensitive to variations along θ .

$$\text{TV}_{\alpha,\theta}(f) = \sum_{i,j} \sup_{t \in E_{\alpha,\theta}} \langle \Delta f(i,j), t \rangle$$

Directional TV

- ▶ Consider an image with a dominant direction. The $TV_{\alpha,\theta}$ with $\alpha = 3$ is:



- ▶ It is possible to use directional TV as a prior for images with a dominant direction.

Image Denoising with Directional TV

- ▶ Regular TV is used in image denoising as

$$f^* = \operatorname{argmin}_f \frac{1}{2} \|y - f\|_2^2 + \lambda TV(f)$$

- ▶ Replace $TV(f)$ with $TV_{\alpha,\theta}(f)$

$$f^* = \operatorname{argmin}_f \frac{1}{2} \|y - f\|_2^2 + \lambda TV_{\alpha,\theta}(f)$$

- ▶ Directional TV forces solutions to have less variations in the chosen direction θ .
- ▶ Amount of forcing can be adjusted with α .

Implementation of Directional TV for Image Denoising

Cost function

$$f^* = \operatorname{argmin}_f \frac{1}{2} \|y - f\|_2^2 + \lambda \operatorname{TV}_{\alpha, \theta}(f)$$

If

$$\operatorname{TV}_{\alpha, \theta}(f) = \sup_{v(i,j) \in E_{\alpha, \theta}} \langle \Delta f, v \rangle$$

Proposition

For

$$v^* = \operatorname{argmin}_{v(i,j) \in E_{\alpha, \theta}} \|f - \lambda \Delta^T v\|_2^2,$$

set $P_f = \lambda \Delta^T v^*$. Then, $f - P_f$ minimizes the cost function.

Implementation of Directional TV for Image Denoising

- ▶ If we define the rotation and scaling matrices R_θ , Λ_α as,

$$R_\theta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}, \quad \Lambda_\alpha = \begin{bmatrix} \alpha & 0 \\ 0 & 1 \end{bmatrix}.$$

- ▶ Using these, $E_{\alpha,\theta}$ and B_2 are related as $E_{\alpha,\theta} = R_\theta \Lambda_\alpha B_2$.
- ▶ Note that $\mathbf{R}_\theta^T = \mathbf{R}_{-\theta}$ and $\mathbf{\Lambda}_\alpha^T = \mathbf{\Lambda}_\alpha$.
- ▶ $\text{TV}_{\alpha,\theta}$ becomes

$$\begin{aligned} \text{TV}_{\alpha,\theta} &= \sup_{v(i,j) \in R_\theta \Lambda_\alpha B_2} \langle \Delta f, v \rangle \\ &= \sup_{v(i,j) \in B_2} \langle \Delta f, \mathbf{R}_\theta \mathbf{\Lambda}_\alpha v \rangle \\ &= \sup_{v(i,j) \in B_2} \langle \mathbf{\Lambda}_\alpha \mathbf{R}_{-\theta} \Delta f, v \rangle. \end{aligned}$$

Implementation of Directional TV for Image Denoising

Corollary

For

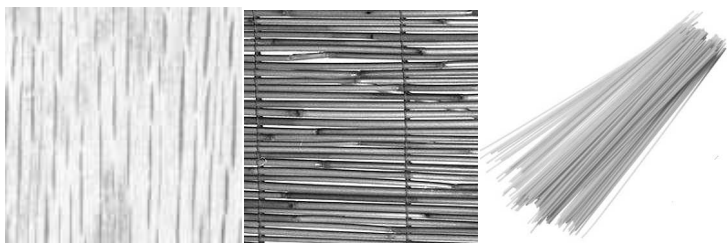
$$v^* = \operatorname{argmin}_{v(i,j) \in B_2} \|f - \lambda \Delta^T \mathbf{R}_\theta \mathbf{\Lambda}_\alpha v\|_2^2,$$

set $P_f = \lambda \Delta^T \mathbf{R}_\theta \mathbf{\Lambda}_\alpha v^*$. Then, $f - P_f$ minimizes the cost function.

- ▶ Further details of the algorithm are given in the paper.
- ▶ Matlab code for the implementation can be found at <http://web.itu.edu.tr/ibayram/DTV/>.

Image Denoising Experimental Setup

- ▶ Images used for denoising experiments

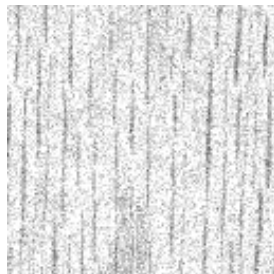


- ▶ Image pixel values are normalized to $[0,1]$.
- ▶ Iid Gaussian noise with $\sigma = 0.1$ is added to the images.
- ▶ Images are denoised using regular TV and directional TV.
- ▶ TV parameter λ is chosen to minimize the RMSE of the denoised image.
- ▶ Parameters of directional TV (α, θ) are chosen manually.
- ▶ RMSE is used to compare the denoising results.

Results – Regular vs. Directional TV



(a) Original



(b) Noisy – RMSE=0.1009

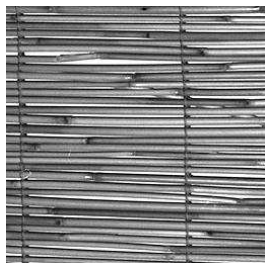


(c) TV – RMSE=0.0489

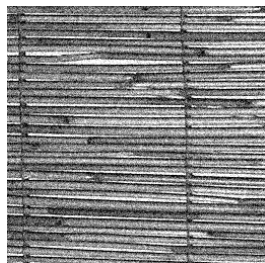


(d) TV_{5, π/2} – RMSE=0.0429

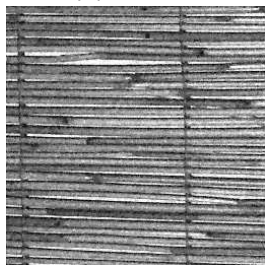
Results – Regular vs. Directional TV



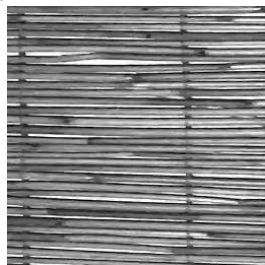
(a) Original



(b) Noisy – RMSE=0.1005

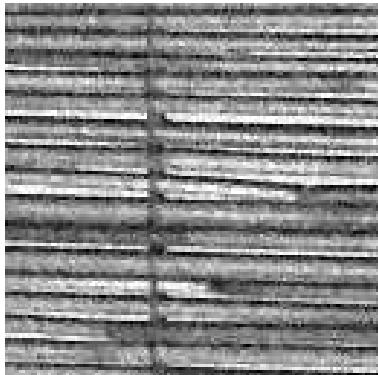


(c) TV – RMSE=0.0354

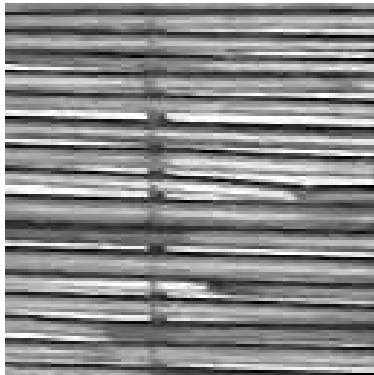


(d) TV_{5,0} – RMSE=0.0279

Results – Regular vs. Directional TV

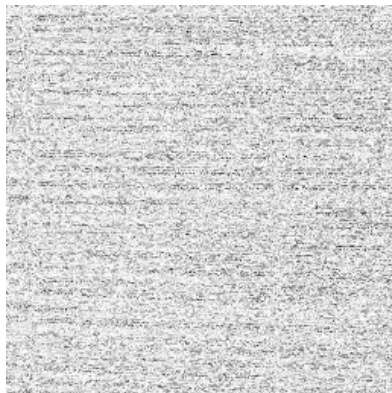


(a) TV

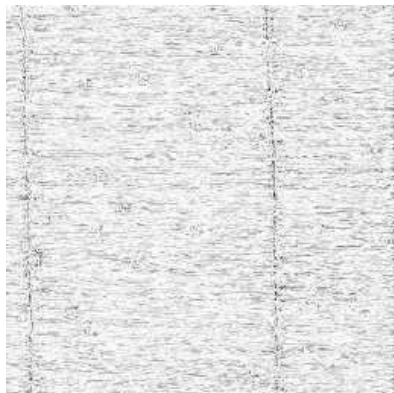


(b) TV_{5,0}

Results – Regular vs. Directional TV



(a) Difference TV

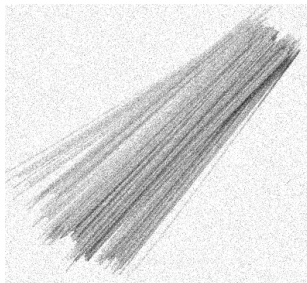


(b) Difference TV_{5,0}

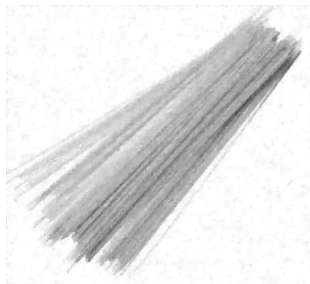
Results – Regular vs. Directional TV



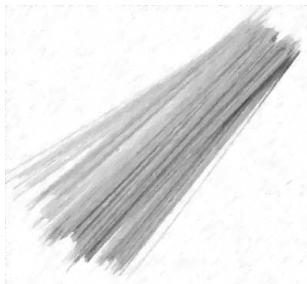
(a) Original



(b) Noisy – (RMSE=0.1002)

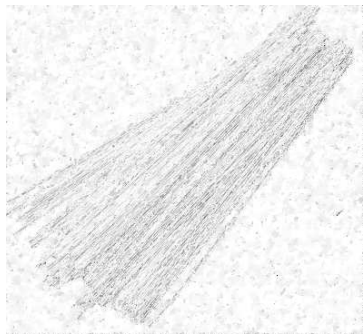


(c) TV – (RMSE=0.0431)

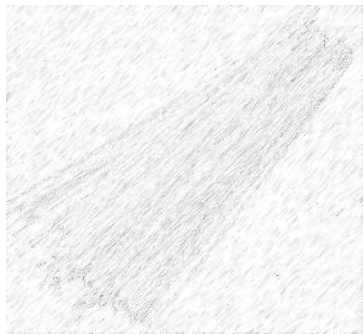


(d) $TV_{5, \pi/4}$ – (RMSE=0.0269)

Results – Regular vs. Directional TV



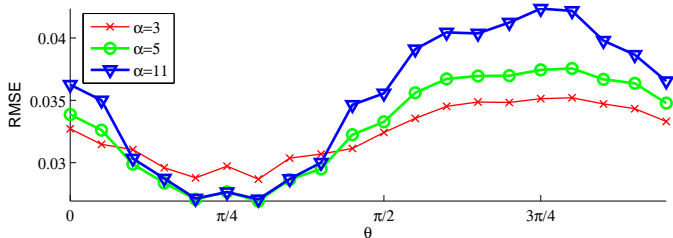
(a) Difference (TV)



(b) Difference ($TV_{5,\pi/4}$)

Effect of Parameter Selection for Directional TV

- ▶ Different parameters of directional TV (α, θ) are used to denoise the image.
- ▶ RMSE values of the denoised images are



- ▶ If α is chosen small, it is not effective.
- ▶ If α is chosen to be large number, RMSE is high if θ is not correctly chosen.

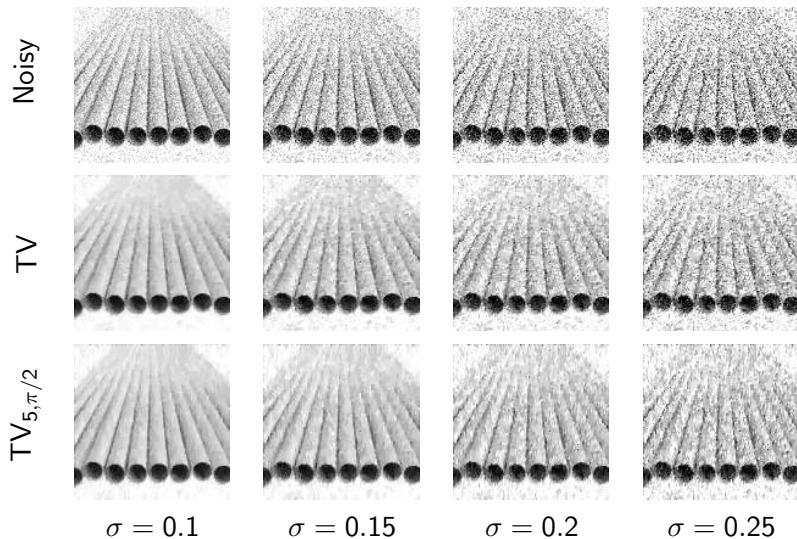
Performance of TV vs Directional TV at Different Noise Levels

- ▶ Denoising performance of regular TV vs directional TV is investigated at different noise levels.
- ▶ Noise with $\sigma = \{0.01, 0.02, \dots, 0.25\}$ are added to the pipe image.

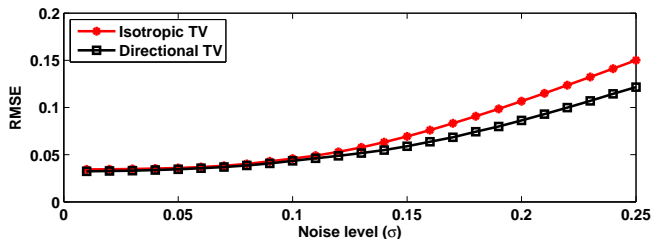


- ▶ Noisy images are denoised with regular TV and directional TV.

Performance of TV vs Directional TV at Different Noise Levels



Performance of TV vs Directional TV at Different Noise Levels



- ▶ RMSE values for denoised pipes images obtained with directional TV with different α and θ parameters.
- ▶ As the noise level increases, denoising with directional TV prior outperforms the regular TV in terms of RMSE.
- ▶ The significance of the prior term increases with the level of noise.

Conclusions

- ▶ A directional TV and its implementation is described.
- ▶ Image denoising with directional TV prior outperforms regular TV in terms of RMSE on images (with a dominant direction).
- ▶ Parameters of directional TV (α, θ) has to be chosen correctly. Otherwise, the denoising fails.
- ▶ The denoising performance between directional TV and regular TV increases with level of noise.

Future Work

- ▶ Parameters of directional TV (that characterizes the ellipse) is same for all pixels.
- ▶ Limitation: Directional TV works on images with a dominant direction.
- ▶ Parameters (α, θ) can be changed for each pixel (requires edge detection).
- ▶ With such a modification, directional TV can be used for other images.
- ▶ Applications of directional TV can be extended to other applications such as sparse sample reconstruction, deconvolution etc.

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