## De-convolution and De-nosing by TV-L1/L2 Regularization with Iterative Gradient Descent Method

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## 1 Deblurring by Total Variation Regularization

Blurry and Gaussian Noisy Image:

$$y = k * u + n = Ku + n \tag{1}$$

$$u^* \in \underset{u}{\operatorname{arg\,min}} ||y - Ku||^2 + \lambda J(u) \tag{2}$$

Total Variation Regularization:

$$J(u) = \sum ||\nabla u|| \tag{3}$$

or

$$J(u) = \sum \sqrt{||\nabla u||^2 + \epsilon} \tag{4}$$

## 2 Solver: Iterative Gradient Descent Method

Gradient Descent:

$$u^{(i+1)} = u^{(i)} - \tau \left( k * (k * u^{(i)} - y) + \lambda GradJ(u^{(i)}) \right)$$
(5)

Where,

- The step size  $\tau$  can be estimated by line search.
- The gradient of the TV term is:

$$GradJ(u) = -div\left(\frac{\nabla u}{\sqrt{||\nabla u||^2 + \epsilon}}\right)$$
 (6)

- The gradient of u can be computed as

$$\nabla u = [u_x, u_y]$$

$$u_x = u_{i+1,j} - u_{i,j}$$

$$u_y = u_{i,j+1} - u_{i,j}$$

$$(7)$$

– Euclidean Norm:

$$||\nabla u|| = \sqrt{u_x^2 + u_y^2} \tag{8}$$

- The divergence of the field is the following scheme:

$$div(u) = \frac{\partial u_x}{\partial x} + \frac{\partial u_y}{\partial y}$$

$$(div(u))_{i,j} = \begin{cases} u_x(i,j) - u_x(i-1,j), & \text{if } 1 < i < n; \\ u_x(i,j), & \text{if } i = 1; \\ -u_x(i-1,j), & \text{if } i = n. \end{cases}$$

$$+ \begin{cases} u_y(i,j) - u_y(i,j-1), & \text{if } 1 < j < n; \\ u_y(i,j), & \text{if } j = 1; \\ -u_y(i,j-1), & \text{if } j = n. \end{cases}$$

$$(9)$$