

Lecture 18:

Image denoising by solving PDE (derived from energy minimisation problem)

Consider the harmonic - L2 minimization model:

$$\text{minimize } \bar{E}(f) = \int_{\Omega} (f(x, y) - \underbrace{g(x, y)}_{\substack{\text{image domain} \\ \text{Observed}}})^2 dx dy + \int |\nabla f|^2 dx dy$$

(Look for (continuous) image f) Smoothness of f

We find f that minimizes $E(f)$.

Take any function $v(x, y)$. Consider a real-valued function $S: \mathbb{R} \rightarrow \mathbb{R}$:

$$S(\varepsilon) := \bar{E}(f + \varepsilon v) = \int_{\Omega} (f(x, y) + \varepsilon v(x, y) - g(x, y))^2 dx dy + \int |\nabla f + \varepsilon \nabla v|^2 dx dy$$

$$\frac{d}{d\varepsilon} S(\varepsilon) = 2 \int_{\Omega} (f(x, y) + \varepsilon v(x, y) - g(x, y))^2 dx dy + 2 \int_{\Omega} \left[\left(\frac{\partial f}{\partial x} + \varepsilon \frac{\partial v}{\partial x} \right) \frac{\partial v}{\partial x} + \left(\frac{\partial f}{\partial y} + \varepsilon \frac{\partial v}{\partial y} \right) \frac{\partial v}{\partial y} \right] dx dy$$

If f is a minimizer, $\frac{d}{d\varepsilon} S(\varepsilon) = 0$ for all v ($\because S(0)$ is the minimum)

$$\therefore S'(0) = 0 = 2 \int_{\Omega} (f(x,y) - g(x,y)) v(x,y) dx dy + 2 \int_{\Omega} (f_x v_x + f_y v_y) dx dy \text{ for all } v.$$

Remark: If we can formulate the above equation as follows:

$$\int_{\Omega} T(x,y) v(x,y) dx dy = 0 \text{ for all } v(x,y)$$

then, we can conclude $T(x,y) = 0$ in Ω .

In our case, first term is okay!
Second term NOT okay!

Useful Tool: (Integration by part)

$$\int_{\Omega} \nabla f \cdot \nabla g dx dy = - \int_{\Omega} (\nabla \cdot (\nabla f)) g dx dy + \int_{\partial\Omega} g (\nabla f \cdot \vec{n}) ds$$

$$\nabla \cdot (V_1(x,y), V_2(x,y))$$

$$\frac{\partial V_1}{\partial x} + \frac{\partial V_2}{\partial y}$$

where $\vec{n} = (n_1, n_2)$ = outward normal on the boundary.

$$\text{In our case, we get: } 0 = \int_{\Omega} (f - g) v dx dy - \int_{\Omega} (\nabla \cdot \nabla f) v dx dy + \int_{\partial\Omega} (\nabla f \cdot \vec{n}) v ds$$

Overall, we get : $\int_{\Omega} (f - g - \Delta f) v \, dx \, dy - \int_{\partial\Omega} (\nabla f \cdot \vec{n}) v \, ds = 0$ for all v

We conclude : $\begin{cases} f - g - \Delta f = 0 & \text{in } \Omega \\ \nabla f \cdot \vec{n} = 0 & \text{in } \partial\Omega \end{cases}$ (PDE)

Remark: • Anisotropic diffusion is related to minimizing:

$$E(f) = \int_{\Omega} K(x,y) |\nabla f(x,y)|^2 dx dy$$

- Energy minimization approach for solving imaging problem is called the **Variational image processing!**

$$\bar{E}(f + \varepsilon v) = \int K(x,y) |\nabla f + \varepsilon \nabla v|^2 dx dy$$

$$\begin{aligned}\frac{d}{d\varepsilon} \Big|_{\varepsilon=0} E(f + \varepsilon v) &= \int_{\Omega} K(x,y) (\nabla f + \varepsilon \nabla v) \cdot (\nabla f + \varepsilon \nabla v) \\ &= \int_{\Omega} K(x,y) (2 \nabla f \cdot \nabla v) \\ &= - \int_{\partial\Omega} \nabla \cdot (K(x,y) \nabla f) v + \int_{\partial\Omega} (K(x,y) \nabla f \cdot \vec{n}) v\end{aligned}$$

Total variation (TV) denoising (ROF)

Invented by: Rudin, Osher, Fatemi

Motivation: Previous model: $f = g + \Delta f$. Solve for f from noisy g .

Disadvantage: smooth out edge.

Modification: $f = g + \nabla \cdot (K \nabla f)$

K is small on edges!!

Goal: Given a noisy image $g(x,y)$, we look for $f(x,y)$ that solves:

$$f = g + \lambda \frac{\partial}{\partial x} \left(\frac{1}{|\nabla f|(x,y)} \frac{\partial f}{\partial x} \right) + \frac{\partial}{\partial y} \left(\frac{1}{|\nabla f|(x,y)} \frac{\partial f}{\partial y} \right) \quad (*)$$

Remark: Problem arises if $|\nabla f(x,y)|=0$. Take care of it later.

We'll show that $(*)$ must be satisfied by a minimizer of:

$$\mathcal{J}(f) = \frac{1}{2} \int_{\Omega} (f(x,y) - g(x,y))^2 + \underbrace{\lambda \int_{\Omega} |\nabla f(x,y)|}_{\text{constant parameter} > 0} dx dy$$

Same idea: Let $S(\varepsilon) := E(f + \varepsilon v)$

$$= \int_{\Omega} (f + \varepsilon v - g)^2 + \lambda \int_{\Omega} |\nabla f + \varepsilon \nabla v|$$

$$\frac{d}{d\varepsilon} S(\varepsilon) = \left[\int_{\Omega} (f + \varepsilon v - g) v + \lambda \int_{\Omega} \frac{\nabla f \cdot \nabla v + 2\varepsilon \nabla v \cdot \nabla v}{\sqrt{(\nabla f + \varepsilon \nabla v) \cdot (\nabla f + \varepsilon \nabla v)}} \right]$$

If f is a minimizer, $\frac{d}{d\varepsilon} S(\varepsilon) = 0$ for all v .

$$\begin{aligned} \therefore S'(0) &= 0 = \int_{\Omega} (f - g) v + \lambda \int_{\Omega} \frac{\nabla f \cdot \nabla v}{|\nabla f|} \\ &= \int_{\Omega} (f - g) v - \lambda \int_{\Omega} \nabla \cdot \left(\frac{\nabla f}{|\nabla f|} \right) v + \lambda \int_{\partial\Omega} \left(\frac{\nabla f}{|\nabla f|} \cdot \vec{n} \right) v \\ &= \int_{\Omega} \left[(f - g) - \lambda \nabla \cdot \left(\frac{\nabla f}{|\nabla f|} \right) \right] v + \lambda \int_{\partial\Omega} \left(\frac{\nabla f}{|\nabla f|} \cdot \vec{n} \right) v \quad \text{for all } v \end{aligned}$$

We conclude: $(f - g) - \lambda \nabla \cdot \left(\frac{\nabla f}{|\nabla f|} \right) = 0!!$

In the discrete case,

$$J(f) = \frac{1}{2} \sum_{x=1}^N \sum_{y=1}^N (f(x, y) - g(x, y))^2 + \lambda \sum_{x=1}^N \sum_{y=1}^N \sqrt{(f(x+1, y) - f(x, y))^2 + (f(x, y+1) - f(x, y))^2}$$

J can be regarded as a multi-variable function depending on:
 $f(1, 1), f(1, 2), \dots, f(1, N), f(2, 1), \dots, f(2, N), \dots, f(N, N)$.

If f is a minimizer, then $\frac{\partial J}{\partial f(x, y)} = 0$ for all (x, y) .

$$\begin{aligned} \frac{\partial J}{\partial f(x, y)} &= (f(x, y) - g(x, y)) + \lambda \frac{2(f(x+1, y) - f(x, y))(-1) + 2(f(x, y+1) - f(x, y))(-1)}{2\sqrt{(f(x+1, y) - f(x, y))^2 + (f(x, y+1) - f(x, y))^2}} \\ &\quad + \lambda \frac{2(f(x, y) - f(x-1, y))}{2\sqrt{(f(x, y) - f(x-1, y))^2 + (f(x-1, y+1) - f(x-1, y))^2}} \\ &\quad + \lambda \frac{2(f(x, y) - f(x, y-1))}{2\sqrt{(f(x+1, y-1) - f(x, y-1))^2 + (f(x, y) - f(x, y-1))^2}} = 0 \end{aligned}$$

By simplification:

$$\begin{aligned}
 f(x, y) - g(x, y) &= \lambda \left\{ \frac{f(x+1, y) - f(x, y)}{\sqrt{(f(x+1, y) - f(x, y))^2 + (f(x, y+1) - f(x, y))^2}} \right. \\
 &\quad \left. - \frac{f(x, y) - f(x-1, y)}{\sqrt{(f(x, y) - f(x-1, y))^2 + (f(x-1, y+1) - f(x-1, y))^2}} \right\} = \frac{\frac{\partial f}{\partial x}|_{(x, y)}}{|\nabla f|_{(x, y)}} \\
 &\quad + \lambda \left\{ \frac{f(x, y+1) - f(x, y)}{\sqrt{(f(x+1, y) - f(x, y))^2 + (f(x, y+1) - f(x, y))^2}} \right. \\
 &\quad \left. - \frac{f(x, y) - f(x, y-1)}{\sqrt{(f(x+1, y-1) - f(x, y-1))^2 + (f(x, y) - f(x, y-1))^2}} \right\} = \frac{\frac{\partial f}{\partial y}|_{(x, y)}}{|\nabla f|_{(x, y)}}
 \end{aligned}$$

Discretization of

$$f - g = \lambda \nabla \cdot \left(\frac{\nabla f}{\|\nabla f\|} \right)$$

$$\frac{\partial f}{\partial y}(x,y-1)$$