

Lecture 21:

Image processing by minimization
(Variational approach)

- ① Consider a minimization model (usually in continuous sense)
- ② Derive a PDE related to minimization model
- ③ Discretize the PDE to get a linear system.

e.g. Total-variation (TV) denoising model, (ROF)
(Rudin - Osher - Fatemi)

2D integration by part formula

Let $f: [a, b] \times [a, b] \rightarrow \mathbb{R}$ and $g: [a, b] \times [a, b] \rightarrow \mathbb{R}$.

Assume $f(a, y) = f(b, y) = f(x, a) = f(x, b) = 0$.

$g(a, y) = g(b, y) = g(x, a) = g(x, b) = 0$.

Then: $\int_a^b \int_a^b \nabla f(x, y) \cdot \nabla g(x, y) dx dy = - \int_a^b \int_a^b \Delta f(x, y) g(x, y) dx dy$

Proof: $\int_a^b \int_a^b \underbrace{\frac{\partial f}{\partial x} \frac{\partial g}{\partial x} + \frac{\partial f}{\partial y} \frac{\partial g}{\partial y}}_{\nabla f \cdot \nabla g} dx dy = - \int_a^b \int_a^b \left(\frac{\partial^2 f}{\partial x^2} \right) g dx dy + \int_a^b \left(\frac{\partial f}{\partial x} \right) g \Big|_{x=a}^{x=b} dy$

$- \int_a^b \int_a^b \left(\frac{\partial^2 f}{\partial y^2} \right) g dx dy + \int_a^b \frac{\partial f}{\partial y} g \Big|_{y=a}^{y=b} dx$

$= - \int_a^b \int_a^b \underbrace{\left(\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} \right)}_{\Delta f} g dx dy$

Also,

$$\int_a^b \int_a^b \left(k(x,y) \nabla f(x,y) \right) \cdot \nabla g(x,y) \, dx \, dy = - \int_a^b \int_a^b \nabla \cdot \left(k(x,y) \nabla f(x,y) \right) g(x,y) \, dx \, dy$$

where $k: [a,b] \times [a,b] \rightarrow \mathbb{R}$.

Proof:

$$\int_a^b \int_a^b \left[k(x,y) \frac{\partial f}{\partial x} \frac{\partial g}{\partial x} + k(x,y) \frac{\partial f}{\partial y} \frac{\partial g}{\partial y} \right] dx \, dy$$

$$= - \int_a^b \int_a^b \frac{\partial}{\partial x} \left(k(x,y) \frac{\partial f}{\partial x} \right) g \, dx \, dy + \int_a^b \cancel{k(x,y)} \frac{\partial f}{\partial x} g \Big|_{x=a}^{x=b} dy$$

$$- \int_a^b \int_a^b \frac{\partial}{\partial y} \left(k(x,y) \frac{\partial f}{\partial y} \right) g \, dx \, dy + \int_a^b \cancel{k(x,y)} \frac{\partial f}{\partial y} g \Big|_{y=a}^{y=b} dx$$

$$= - \int_a^b \int_a^b \left[\frac{\partial}{\partial x} \left(k(x,y) \frac{\partial f}{\partial x} \right) + \frac{\partial}{\partial y} \left(k(x,y) \frac{\partial f}{\partial y} \right) \right] g \, dx \, dy$$

$\nabla \cdot (k(x,y) \nabla f)$

In general, we have:

Useful Tool: (Integration by part)

$$\nabla \cdot (V_1(x,y), V_2(x,y)) \\ \stackrel{!}{=} \frac{\partial V_1}{\partial x} + \frac{\partial V_2}{\partial y}$$

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

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

or more generally,

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

Another useful fact:

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

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

Example: Suppose we have the following integral equation:

$$\int_a^b \int_a^b (f(x,y) - g(x,y)) v(x,y) dx dy + \int_a^b \int_a^b \nabla \cdot \nabla f(x,y) K(x,y) v(x,y) dx dy = 0$$

for all $v(x,y)$. Then: we have:

$$\int_a^b \int_a^b \left[(f(x,y) - g(x,y)) + K(x,y) \nabla \cdot \nabla f(x,y) \right] v(x,y) dx dy = 0$$

for all $v(x,y)$

We can conclude: $(f(x,y) - g(x,y)) + K(x,y) \nabla \cdot \nabla f(x,y) = 0$
for all $(x,y) \in [a,b] \times [a,b]$

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

Consider the harmonic - L2 minimization model:

$$\text{minimize } \bar{E}(f) = \int_a^b \int_a^b \underbrace{(f(x,y) - g(x,y))^2}_{\text{Observed}} dx dy + \int_a^b \int_a^b \underbrace{|\nabla f|^2}_{\text{Smoothness of } f} dx dy$$

(Look for (continuous) image f)

Assume that $f(x,y) = g(x,y) = 0$ on the boundary of $[a,b] \times [a,b]$.

Suppose f minimizes $E(f)$. Let $v: [a,b] \times [a,b] \rightarrow \mathbb{R}$ such that

$v(x,y) = 0$ on the boundary of $[a,b] \times [a,b]$.

Consider $f^\epsilon = f + \epsilon v: [a,b] \times [a,b] \rightarrow \mathbb{R}$, which is another image with $f^\epsilon(x,y) = 0$ on the boundary of $[a,b] \times [a,b]$.

$$f^\epsilon(x,y) = \underbrace{f(x,y)}_0 + \epsilon \underbrace{v(x,y)}_0 = 0 \text{ on } \partial([a,b] \times [a,b]).$$

Consider $S: \mathbb{R} \rightarrow \mathbb{R}$ defined by:

$$S(\varepsilon) \stackrel{\text{def}}{=} E(f^\varepsilon) = E(f + \varepsilon v).$$

Note that $S(0) = E(f) = \text{minimum of } E$. Thus, S attains its minimum at $\varepsilon = 0$.

$$\therefore \frac{dS}{d\varepsilon}(0) = 0.$$

$$\begin{aligned} \text{Now, } \frac{d}{d\varepsilon} \Big|_{\varepsilon=0} S(\varepsilon) &= \frac{d}{d\varepsilon} \Big|_{\varepsilon=0} E(f + \varepsilon v) = \frac{d}{d\varepsilon} \Big|_{\varepsilon=0} \left(\int_a^b \int_a^b (f(x,y) + \varepsilon v(x,y) - g(x,y))^2 dx dy \right. \\ &\quad \left. + \int_a^b \int_a^b |\nabla(f + \varepsilon v)(x,y)|^2 dx dy \right) \\ &= \int_a^b \int_a^b 2(f(x,y) + \varepsilon v(x,y) - g(x,y)) v(x,y) \Big|_{\varepsilon=0} dx dy \\ &\quad + \int_a^b \int_a^b (2 \nabla f \cdot \nabla v + 2\varepsilon |\nabla v|^2) \Big|_{\varepsilon=0} dx dy \\ &= \int_a^b \int_a^b 2(f(x,y) - g(x,y)) v(x,y) dx dy + \int_a^b \int_a^b 2 \nabla f(x,y) \cdot \nabla v(x,y) dx dy \end{aligned}$$

$\begin{aligned} &= (\nabla f + \varepsilon \nabla v) \cdot (\nabla f + \varepsilon \nabla v) \\ &= \nabla f \cdot \nabla f + 2\varepsilon \nabla f \cdot \nabla v + \varepsilon^2 \nabla v \cdot \nabla v \\ &= |\nabla f|^2 + 2\varepsilon \nabla f \cdot \nabla v + \varepsilon^2 |\nabla v|^2 \end{aligned}$

$$\therefore S'(0) = 0 = 2 \int_a^b \int_a^b (f(x,y) - g(x,y)) v(x,y) dx dy + 2 \int_a^b \int_a^b \left(\frac{\partial f}{\partial x}(x,y) \frac{\partial v}{\partial x}(x,y) + \frac{\partial f}{\partial y}(x,y) \frac{\partial v}{\partial y}(x,y) \right) dx dy$$

for all $v(x,y)$. — (*)

If we can formulate (*) in the form:

$$\int_a^b \int_a^b T(x,y) v(x,y) = 0 \quad \text{for all } v(x,y),$$

then we can conclude that $T(x,y) = 0$ in $[a,b] \times [a,b]$.

Remark: • First term is in the form $\int_a^b \int_a^b T(x,y) v(x,y)$

• Second term is NOT.

Need to reformulate the second term.

Strategy: integration by part.

Second term:
$$\int_a^b \int_a^b \nabla f(x,y) \nabla v(x,y) dx dy = 2 \int_a^b \int_a^b \Delta f(x,y) v(x,y) dx dy.$$

All together, we have

$$0 = S'(0) = \int_a^b \int_a^b 2(f(x,y) - g(x,y)) v(x,y) dx dy - 2 \int_a^b \int_a^b \Delta f(x,y) v(x,y) dx dy$$

$$\therefore \int_a^b \int_a^b \left(2(f(x,y) - g(x,y)) - 2 \Delta f(x,y) \right) v(x,y) dx dy = 0 \text{ for all } v(x,y).$$

We conclude:

$$2(f(x,y) - g(x,y)) - 2 \Delta f(x,y) = 0 \text{ for } (x,y) \in [a,b] \times [a,b]$$

or $f(x,y) - g(x,y) - \Delta f(x,y) = 0$ (converse of Laplacian masking !!)

Remark: More generally, if we do not enforce $f(x,y) = g(x,y) = 0$ on the boundary of the image domain Ω . Then, we do not enforce $v(x,y) = 0$ on $\partial\Omega$.

In this case, we have:

$$\begin{aligned} 0 = S'(0) &= \int_{\Omega} 2(f(x,y) - g(x,y)) v(x,y) \, dx dy + \int_{\Omega} 2 \nabla f(x,y) \cdot \nabla v(x,y) \, dx dy \\ &= \int_{\Omega} 2(f(x,y) - g(x,y)) v(x,y) \, dx dy - \int_{\Omega} 2 \Delta f(x,y) v(x,y) \, dx dy \\ &\quad + \int_{\partial\Omega} (2 \nabla f(x,y) \cdot \vec{n}(x,y)) v(x,y) \, ds \end{aligned}$$

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)

Example: Consider an image denoising model to find $f: \frac{[a,b] \times [a,b]}{\Omega} \rightarrow \mathbb{R}$

that minimizes:

$$E(f) = \int_a^b \int_a^b (f(x,y) - g(x,y))^2 dx dy + \int_a^b \int_a^b |\nabla f(x,y)|^4 dx dy.$$

Suppose f minimizes $E(f)$. Assume $f(x,y) = g(x,y) = 0$ for all $(x,y) \in \partial D$. Find a partial differential equation that f must satisfy.

Solution: Suppose f minimizes $E(f)$. For any $v: D \rightarrow \mathbb{R}$ such

that $v(x,y) = 0$ on ∂D , we have:

$$\left\{ \begin{array}{l} f^\varepsilon \stackrel{\text{def}}{=} f + \varepsilon v \text{ is an image with} \\ f^\varepsilon(x,y) = f(x,y) + \varepsilon v(x,y) = 0 \text{ on } \partial D. \end{array} \right.$$

Consider $S: \mathbb{R} \rightarrow \mathbb{R}$ where $S(\varepsilon) \stackrel{\text{def}}{=} E(f^\varepsilon) = E(f + \varepsilon v)$

Then, $S(0) = E(f) = \text{minimum of } E$. Thus, S attains minimum at $\varepsilon = 0$.

$$\therefore \left. \frac{dS}{d\varepsilon} \right|_{\varepsilon=0} = 0 \text{ for all } v: D \rightarrow \mathbb{R}$$

Now,

$$0 = \left. \frac{dS}{d\varepsilon} \right|_{\varepsilon=0} = \left. \frac{d}{d\varepsilon} \right|_{\varepsilon=0} \left(\int_D (f(x,y) + \varepsilon v(x,y) - g(x,y))^2 dx dy + \int_D |\nabla(f + \varepsilon v)(x,y)|^2 dx dy \right)$$

$(|\nabla f + \varepsilon \nabla v|^2)^2$

$(\nabla f + \varepsilon \nabla v) \cdot (\nabla f + \varepsilon \nabla v)$

$(\nabla f \cdot \nabla f + 2\varepsilon \nabla f \cdot \nabla v + \varepsilon^2 \nabla v \cdot \nabla v)$

$(|\nabla f|^2 + 2\varepsilon \nabla f \cdot \nabla v + \varepsilon^2 |\nabla v|^2)$

$$\begin{aligned} \therefore 0 &= \frac{dS}{d\varepsilon}(0) = \int_D 2(f(x,y) + \varepsilon v(x,y) - g(x,y)) v(x,y) \Big|_{\varepsilon=0} dx dy \\ &\quad + \int_D 2(|\nabla f|^2 + 2\varepsilon \nabla f \cdot \nabla v + \varepsilon^2 |\nabla v|^2) (2 \nabla f \cdot \nabla v + 2\varepsilon |\nabla v|^2) \Big|_{\varepsilon=0} dx dy \end{aligned}$$

$$\Rightarrow 0 = \int_D 2(f(x,y) - g(x,y)) v(x,y) dx dy + \int_D 4(|\nabla f|^2) \nabla f \cdot \nabla v dx dy$$

$$= \int_D 2(f(x,y) - g(x,y)) v(x,y) dx dy - \int_D \left(4 \nabla \cdot (|\nabla f|^2 \nabla f) \right) v(x,y) dx dy$$

$$+ \int_{\partial D} \left(4(|\nabla f|^2) \nabla f(x,y) \cdot \vec{n}(x,y) \right) v(x,y) dx dy$$

$\begin{matrix} 0 \\ 0 \\ 0 \end{matrix}$

All together, we have:

$$0 = \int_D \left(2(f(x,y) - g(x,y)) - 4 \nabla \cdot (|\nabla f(x,y)|^2 \nabla f(x,y)) \right) v(x,y) dx dy$$

for all $v(x,y)$.

We can conclude that:

$$f(x,y) - g(x,y) - 4 \nabla \cdot (|\nabla f(x,y)|^2 \nabla f(x,y)) = 0 \text{ in } D.$$

(Partial differential equation)

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!**

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

$$\frac{d}{d\varepsilon} \Big|_{\varepsilon=0} E(f + \varepsilon v) = \frac{d}{d\varepsilon} \int k(x,y) (\nabla f + \varepsilon \nabla v) \cdot (\nabla f + \varepsilon \nabla v)$$

$$= \int_{\Omega} k(x,y) (2 \nabla f \cdot \nabla v)$$

$$= - \int_{\Omega} \nabla \cdot (k(x,y) \nabla f) v + \int_{\partial \Omega} (k(x,y) \nabla f \cdot \vec{n}) v$$