

## Final Exam:

Topics: Fitting up to Energy Minimization,  
NOT including Segmentation.  
i.e. the topics in Final Practice.

Since DFI is highly related to the topics,  
You should also do some revision on DFI.

$$f = (f_{ij})_{0 \leq i, j \in \mathbb{N}-1} \in \mathbb{R}^{N \times N}$$

$$G_x(f)(x, y) = \frac{1}{4} f(x+1, y) + \frac{1}{2} f(x, y) + \frac{1}{4} f(x-1, y)$$

$$G_y(f)(x, y) = \frac{1}{4} f(x, y+1) + \frac{1}{2} f(x, y) + \frac{1}{4} f(x, y-1)$$

$G_x, G_y$  linear and Shift / Position invariant

$\Rightarrow$  both are convolution.

$$G_x \circ G_y(f) = G_x(G_y(f))$$

$$G_x \circ G_y(f)(x, y)$$

$$= \frac{1}{4} G_y(f)(x+1, y) + \frac{1}{2} G_y(f)(x, y) + \frac{1}{4} G_y(f)(x-1, y)$$

$$= \frac{1}{4} \left( \frac{1}{4} f(x+1, y+1) + \frac{1}{2} f(x+1, y) + \frac{1}{4} f(x+1, y-1) \right) \\ + \frac{1}{2} \left( \frac{1}{4} f(x, y+1) + \frac{1}{2} f(x, y) + \frac{1}{4} f(x, y-1) \right)$$

$$+ \frac{1}{4} \left( \frac{1}{4} f(x-1, y+1) + \frac{1}{2} f(x-1, y) + \frac{1}{4} f(x-1, y-1) \right)$$

$$= \frac{1}{16} f(x-1, y-1) + \frac{1}{8} f(x-1, y) + \frac{1}{16} f(x-1, y+1)$$

$$+ \frac{1}{8} f(x, y-1) + \frac{1}{4} f(x, y) + \frac{1}{8} f(x, y+1)$$

$$+ \frac{1}{16} f(x+1, y-1) + \frac{1}{8} f(x+1, y) + \frac{1}{16} f(x+1, y+1)$$

$$\therefore G_x \circ G_y (f) = h * f$$

$$= \begin{bmatrix} 1/4 & 1/8 & 0 & \dots & 0 & 1/8 \\ 1/8 & 1/16 & 0 & \dots & 0 & 1/16 \\ 0 & 0 & & & & \\ \vdots & \vdots & & & & \\ 1/8 & 1/16 & 0 & \dots & 0 & 1/16 \end{bmatrix} * f.$$

In General, convolution is associative,

$$\text{i.e. } h_1 * (h_2 * f) = (h_1 * h_2) * f$$

$$\text{Note : } G_x (f) = \begin{bmatrix} 1/2 & 0 & \dots & 0 \\ 1/4 & 0 & \dots & 0 \\ 0 & \vdots & \ddots & \vdots \\ 0 & & & \\ 1/4 & 0 & \dots & 0 \end{bmatrix} * f$$

$$G_y (f) = \begin{bmatrix} 1/2 & 1/4 & 0 & \dots & 0 & 1/4 \\ 0 & 0 & & & 0 & \\ \vdots & \vdots & & & & \\ 0 & 0 & & & & 0 \end{bmatrix} * f$$

$$\begin{aligned} G_x \circ G_y (f) &= h_1 * (h_2 * f) \\ &= (h_1 * h_2) * f \end{aligned}$$

$$h_1 * h_2 = \begin{bmatrix} 1/4 & 1/8 & 0 & \dots & 0 & 1/8 \\ 1/8 & 1/16 & 0 & \dots & 0 & 1/16 \\ \vdots & \vdots & & & & \vdots \\ 0 & 0 & & & & \\ 1/8 & 1/16 & 0 & \dots & 0 & 1/16 \end{bmatrix}$$

Suppose  $f, g \equiv 0$  on  $\partial D$

Consider the Energy:

$$\bar{E}(f) = \int_D (f(x, y) - g(x, y))^2 + \int |\nabla f|^2$$

Let  $f$  minimizer of  $\bar{E}$ ,  $v \equiv 0$  on  $D$ ,

$$S(\varepsilon) = \bar{E}(f + \varepsilon v)$$

$$\begin{aligned} \frac{d}{d\varepsilon} S(\varepsilon) &= \frac{d}{d\varepsilon} \int_D (f - g + \varepsilon v)^2 \\ &\quad + \frac{d}{d\varepsilon} \int_D \langle \nabla f + \varepsilon \nabla v, \nabla f + \varepsilon \nabla v \rangle \\ &= \frac{d}{d\varepsilon} \int_D (f - g + \varepsilon v)^2 \\ &\quad + \frac{d}{d\varepsilon} \int_D |\nabla f|^2 + 2\varepsilon \nabla f \cdot \nabla v + \varepsilon^2 |\nabla v|^2 \end{aligned}$$

$$= 2 \int_D (f - g + \varepsilon v) v$$

$$+ 2 \int_D \nabla f \cdot \nabla v + 2\varepsilon |\nabla v|^2$$

$$f \text{ minimizer} \Rightarrow \left. \frac{d}{d\varepsilon} S(\varepsilon) \right|_{\varepsilon=0} = 0$$

$$\Rightarrow 0 = \int_D (f - g) v + \int_D \nabla f \cdot \nabla v$$

By Divergence Thm (Integration by Parts),

$$\int_D \nabla f \cdot \nabla v = \int_{\partial D} v (\nabla f \cdot \vec{n}) - \int_D v \Delta f$$

By assumption,  $v \equiv 0$  on  $\partial D$ ,

$$\int_D \nabla f \cdot \nabla v = - \int_D v \Delta f$$

Then,

$$0 = \int_D (f - g - \Delta f) v$$

$v$  is arbitrary,

$$\Rightarrow \begin{cases} 0 = f - g - \Delta f \\ f = g = 0 \text{ on } \partial D \end{cases}$$

To do:

$$\text{Solve } \begin{cases} 0 = f - g - \Delta f \\ f = g = 0 \text{ on } \partial D \end{cases} \text{ numerically.}$$

Now,  $f, g$  are images  $\in \mathbb{R}^{N \times N}$

$$\Delta f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} \text{ is continuous differential operator,}$$

we have to discretized it such that it can be applied to matrix.

Taylor Expansion:

$$(1) - f(x+h) = f(x) + f'(x)h + \frac{f''(x)}{2}h^2 + \dots$$

$$(2) - f(x-h) = f(x) - f'(x)h + \frac{f''(x)}{2}h^2 + \dots$$

(1) + (2) :

$$f(x+h) + f(x-h) = 2f(x) + f''(x)h^2 + \text{higher order terms.}$$

$$\Rightarrow f''(x) \approx \frac{f(x+h) + f(x-h) - 2f(x)}{h^2}$$

In matrix, we set  $h=1$ , approximate  $\frac{\partial^2 f}{\partial x^2}$  by

$$\frac{\partial^2 f}{\partial x^2}(x, y) \approx f(x+1, y) + f(x-1, y) - 2f(x, y)$$

$$\therefore \Delta f(x, y) \approx -4f(x, y) + f(x+1, y) + f(x-1, y) \\ + f(x, y+1) + f(x, y-1)$$

So we write the PDE as:

find  $f \in \mathbb{R}^{N \times N}$  s.t.

$$0 = f(x, y) - \gamma(x, y) -$$

$$\left[ -4f(x, y) + f(x+1, y) + f(x-1, y) \right. \\ \left. + f(x, y+1) + f(x, y-1) \right]$$

$$\Leftrightarrow \gamma(x, y) = 5f(x, y) - f(x+1, y) - f(x-1, y) \\ - f(x, y+1) - f(x, y-1) \quad - (*)$$

Write  $\vec{\gamma}$  be the column vector form of  $\gamma$ ,

$\vec{f}$  be the column vector form of  $f$ ,

write RHS as  $\mathcal{O}(\vec{f})$

Note  $\mathcal{O}$  is linear :

$$\mathcal{O}(\alpha \vec{f}_1 + \beta \vec{f}_2) = \alpha \mathcal{O}(\vec{f}_1) + \beta \mathcal{O}(\vec{f}_2)$$

$\therefore \mathcal{O}$  is some linear transform

and we can write (\*)

$$\text{as } \vec{g} = A \vec{f}, \quad A \in \mathbb{R}^{N^2 \times N^2} \quad \text{some matrix}$$

$$f = \begin{bmatrix} | & | & & | \\ \vec{f}_0 & \vec{f}_1 & \dots & \vec{f}_{N-1} \\ | & | & & | \end{bmatrix}, \quad \vec{f} = \begin{bmatrix} \vec{f}_0 \\ \vec{f}_1 \\ \vdots \\ \vec{f}_{N-1} \end{bmatrix}$$

$$0 \leq j \leq N^2 - 1, \quad ,$$

$$\text{Write } j = m_j N + n_j, \quad m_j = \lfloor \frac{j}{N} \rfloor, \quad n_j = j \bmod N$$

$$\text{Note } 0 \leq m_j, n_j \leq N-1$$

Then

$$\vec{f}(j) = f(m_j, n_j)$$



$$(A \vec{f})_i = \sum_{j=0}^{N^2-1} A(i,j) \vec{f}(j)$$

$$\Leftrightarrow \mathcal{O}(f)(m_i, n_i) = \sum_{j=0}^{N^2-1} A(i,j) f(m_j, n_j)$$

$$\begin{aligned} \Leftrightarrow & \mathcal{S} f(m_i, n_i) \\ & - f(m_{i+1}, n_i) \\ & - f(m_{i-1}, n_i) \\ & - f(m_i, n_{i+1}) \\ & - f(m_i, n_{i-1}) \end{aligned} = \sum_{j=0}^{N^2-1} A(i,j) f(m_j, n_j)$$

By periodicity,

$$\bullet = \mathcal{S} f(m_i, n_i) - f(m_i^+, n_i) - f(m_i^-, n_i) - f(m_i, n_i^+) - f(m_i, n_i^-)$$

$$\text{where } m_i^+ = \begin{cases} 0, & \text{if } m_{i+1} = 0 \\ m_{i+1}, & \text{else} \end{cases}, \quad m_i^- = \begin{cases} N-1, & \text{if } m_{i-1} = -1 \\ m_{i-1}, & \text{else} \end{cases}$$

$$n_i^+ = \begin{cases} 0, & \text{if } n_{i+1} = 0 \\ n_{i+1}, & \text{else} \end{cases}, \quad n_i^- = \begin{cases} N-1, & \text{if } n_{i-1} = -1 \\ n_{i-1}, & \text{else} \end{cases}$$

$$\begin{aligned} \bullet & = A(i, i) f(m_i, n_i) \quad \text{where } i^+ = m_i^+ N + n_j \\ & + A(i^+, i) f(m_i^+, n_j) \quad i^- = m_i^- N + n_j \\ & + A(i^-, i) f(m_i^-, n_j) \quad i_+ = m_i N + n_i^+ \\ & + A(i, i_+) f(m_i, n_i^+) \quad i_- = m_i N + n_i^- \\ & + A(i, i_-) f(m_i, n_i^-) \\ & + \text{the rest.} \end{aligned}$$

$$\begin{aligned}
 \Rightarrow \quad A(i, i) &= 5, \\
 A(i^+, i) &= -1, \\
 A(i^-, i) &= -1, \\
 A(i, i^+) &= -1, \\
 A(i, i^-) &= -1, \quad \forall 0 \leq i \leq N^2 - 1
 \end{aligned}$$

the rest are zero.