

MATH3360: Mathematical Imaging

Assignment 4 Solutions

1. (a) D is block-circulant, i.e.

$$D = \begin{pmatrix} D_0 & D_{N-1} & D_{N-2} & \cdots & D_1 \\ D_1 & D_0 & D_{N-1} & \cdots & D_2 \\ D_2 & D_1 & D_0 & \cdots & D_{2,N-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ D_{N-1} & D_{N-2} & D_{N-3} & \cdots & D_0 \end{pmatrix},$$

with each $N \times N$ block D_0, D_1, \dots, D_{N-1} being circulant. Hence denoting the (k, l) -th block of a $N \times N$ matrix A of $N \times N$ blocks by $A_{k,l}$, we have:

$$\begin{aligned} (W^{-1}DW)_{k,l} &= \sum_{m=0}^{N-1} W_{k,m}^{-1} (DW)_{m,l} \\ &= \sum_{m=0}^{N-1} W_{k,m}^{-1} \sum_{n=0}^{N-1} D_{m,n} W_{n,l} \\ &= \sum_{m,n=0}^{N-1} \overline{W_N(k,m)} W_N(n,l) \overline{W_N} D_{m,n} W_N \\ &= \frac{1}{N} \sum_{m,n=0}^{N-1} e^{2\pi j \frac{ln-km}{N}} \overline{W_N} D_{m,n} W_N, \end{aligned}$$

$$\begin{aligned} \text{and thus } (W^{-1}DW)_{k,l}(q,r) &= \frac{1}{N} \sum_{m,n=0}^{N-1} e^{2\pi j \frac{ln-km}{N}} \sum_{s=0}^{N-1} \overline{W_N(q,s)} (D_{m,n} W_N)(s,r) \\ &= \frac{1}{N\sqrt{N}} \sum_{m,n,s=0}^{N-1} e^{2\pi j \frac{ln-km-qs}{N}} \sum_{t=0}^{N-1} D_{m,n}(s,t) W_N(t,r) \\ &= \frac{1}{N^2} \sum_{m,n,s,t=0}^{N-1} e^{2\pi j \frac{ln+rt-km-qs}{N}} D_{m-n,0}(s-t,0) \\ &= \frac{1}{N^2} \sum_{m,s=0}^{N-1} \sum_{n'=m-N+1}^m \sum_{t'=r-N+1}^s e^{2\pi j \frac{l(m-n')+r(s-t')-km-qs}{N}} D_{n',0}(t',0) \end{aligned}$$

$$\begin{aligned} &= \frac{1}{N^2} \sum_{m,s=0}^{N-1} e^{2\pi j \frac{(l-k)m+(r-q)s}{N}} \sum_{n',t'=0}^{N-1} e^{2\pi j \frac{ln'+rt'}{N}} D(t'+n'N,0) \\ &= \frac{1}{N^2} \cdot N\delta(l-k) \cdot N\delta(r-q) \cdot \sum_{n',t'=0}^{N-1} e^{-2\pi j \frac{ln'+rt'}{N}} h(t',n') \\ &= N^2 \text{DFT}(h)(r,l)\delta(l-k)\delta(r-q). \end{aligned}$$

Hence $W^{-1}DW$ is diagonal.

Thus the eigenvalues of D are the diagonal entries of $W^{-1}DW$, i.e. $\{N^2 DFT(h)(u, v) : 0 \leq u, v \leq N-1\}$.

- (b) Since L is also block-circulant, it is also diagonalizable by W . Denote $W^{-1}DW$ by Λ_D and $W^{-1}LW$ by Λ_L . Then

$$\begin{aligned} \lambda D^T D + L^T L &= \lambda D^* D + L^* L \\ &= \lambda (W \Lambda_D W^{-1})^* (W \Lambda_D W^{-1}) + (W \Lambda_L W^{-1})^* (W \Lambda_L W^{-1}) \\ &= \lambda W \Lambda_D^* \Lambda_D W^{-1} + W \Lambda_L^* \Lambda_L W^{-1} \end{aligned}$$

and $\lambda D^T = \lambda D^* = \lambda (W \Lambda_D W^{-1})^* = \lambda W \Lambda_D^* W^{-1}$. Hence

$$\begin{aligned} (\lambda W \Lambda_D^* \Lambda_D W^{-1} + W \Lambda_L^* \Lambda_L W^{-1}) \mathcal{S}(f) &= \lambda W \Lambda_D^* W^{-1} \mathcal{S}(g), \\ (\lambda \Lambda_D^* \Lambda_D + \Lambda_L^* \Lambda_L) \mathcal{S}(DFT(f)) &= \lambda \Lambda_D^* \mathcal{S}(DFT(g)) \end{aligned}$$

and thus

$$\begin{aligned} DFT(f)(u, v) &= \frac{\lambda N^2 \overline{DFT(h)(u, v)}}{N^4 [\lambda |DFT(h)(u, v)|^2 + |DFT(p)(u, v)|^2]} DFT(g)(u, v) \\ &= \frac{1}{N^2} \frac{\lambda \overline{DFT(h)(u, v)}}{\lambda |DFT(h)(u, v)|^2 + |DFT(p)(u, v)|^2} DFT(g)(u, v). \end{aligned}$$

2. (a) $E(f) = \int_{\Omega} |f(x, y) - g(x, y) + \epsilon| + \lambda \|\nabla f(x, y)\|^2 dx dy$. Then for any $\varphi : \Omega \rightarrow \mathbb{R}$,

$$\begin{aligned} \left. \frac{\partial E(f + t\varphi)}{\partial t} \right|_{t=0} &= \left. \frac{\partial}{\partial t} \right|_{t=0} \int_{\Omega} |f + t\varphi - g + \epsilon| + \lambda \|\nabla(f + t\varphi)\|^2 dx dy \\ &= \int_{\Omega} \frac{1}{|f - g + \epsilon|} \varphi(f - g + \epsilon) + 2\lambda \nabla f \cdot \nabla \varphi dx dy \\ &= \int_{\Omega} \frac{1}{|f - g + \epsilon|} \varphi(f - g + \epsilon) - 2\lambda \varphi \Delta f dx dy + \int_{\partial\Omega} 2\lambda \varphi \nabla f \cdot \vec{n} d\sigma \end{aligned}$$

So, the minimizer of the energy must satisfy the following PDEs

$$\begin{cases} \frac{1}{|f(x, y) - g(x, y) + \epsilon|} (f(x, y) - g(x, y) + \epsilon) - 2\lambda \Delta f(x, y) = 0 & \text{if } (x, y) \in \Omega \\ \nabla f(x, y) \cdot \vec{n}(x, y) = 0 & \text{if } (x, y) \in \partial\Omega \end{cases}$$

- (b) The iterative scheme in the continuous case is

$$f^{n+1}(x, y) = f^n(x, y) + t \left(- \frac{1}{|f^n(x, y) - g(x, y) + \epsilon|} (f^n(x, y) - g(x, y) + \epsilon) + 2\lambda \Delta f^n(x, y) \right)$$

for some prescribed small positive real number t . Numerically, we can approximate the Laplacian using the following formula

$$\Delta f^n(x, y) dt \approx f^n(x+1, y) + f^n(x-1, y) + f^n(x, y+1) + f^n(x, y-1) - 4f^n(x, y)$$

3. $E(f) = \int_{\Omega} |f(x, y) - g_1(x, y)|^2 + |f(x, y) - g_2(x, y)|^2 + \lambda \|\nabla f(x, y)\|^4 dx dy$. Then for any $\varphi : \Omega \rightarrow \mathbb{R}$,

$$\begin{aligned} \frac{\partial E(f + t\varphi)}{\partial t} \Big|_{t=0} &= \frac{\partial}{\partial t} \Big|_{t=0} \int_{\Omega} |f + t\varphi - g_1|^2 + |f + t\varphi - g_2|^2 + \lambda \|\nabla(f + t\varphi)\|^4 dx dy \\ &= \frac{\partial}{\partial t} \Big|_{t=0} \int_{\Omega} |f + t\varphi - g_1|^2 + |f + t\varphi - g_2|^2 + \lambda \left(\|\nabla(f + t\varphi)\|^2 \right)^2 dx dy \\ &= \int_{\Omega} 2\varphi(f - g_1) + 2\varphi(f - g_2) + 4\lambda \|\nabla f\|^2 \nabla f \cdot \nabla \varphi dx dy \\ &= 2 \int_{\Omega} \varphi(2f - g_1 - g_2) - 2\lambda \varphi \nabla \cdot (\|\nabla f\|^2 \nabla f) dx dy + \int_{\partial\Omega} 2\lambda \varphi \|\nabla f\|^2 \nabla f \cdot \vec{n} d\sigma \end{aligned}$$

Ignoring the boundary condition, the iterative scheme in the continuous case is

$$f^{n+1}(x, y) = f^n(x, y) + t \left(g_1(x, y) + g_2(x, y) - 2f(x, y) + 2\lambda \nabla \cdot (\|\nabla f\|^2 \nabla f)(x, y) \right)$$

where t is a small prescribed positive real number.

4. (a) For any $\varphi : [0, 2\pi] \rightarrow D$,

$$\begin{aligned} &\frac{\partial E_{snake,2}(\gamma + t\varphi)}{\partial t} \Big|_{t=0} \\ &= \frac{\partial}{\partial t} \Big|_{t=0} \left\{ \frac{1}{2} \int_0^{2\pi} \|(\gamma + t\varphi)'(s)\|^2 ds + \frac{\alpha}{2} \int_0^{2\pi} \|(\gamma + t\varphi)''(s)\|^2 ds + \beta \int_0^{2\pi} V((\gamma + t\varphi)(s)) ds \right\} \\ &= \int_0^{2\pi} \gamma'(s)\varphi'(s) ds + \alpha \int_0^{2\pi} \gamma''(s)\varphi''(s) ds + \beta \int_0^{2\pi} \nabla V(\gamma(s))\varphi(s) ds \\ &= \gamma'(s)\varphi(s) \Big|_{s=0}^{2\pi} - \int_0^{2\pi} \gamma''(s)\varphi(s) ds + \alpha \gamma''(s)\varphi'(s) \Big|_{s=0}^{2\pi} - \alpha \int_0^{2\pi} \gamma'''(s)\varphi(s) ds \\ &\quad + \beta \int_0^{2\pi} \nabla V(\gamma(s))\varphi(s) ds \\ &= \int_0^{2\pi} \varphi(s) [\alpha \gamma^{(4)}(s) + \beta \nabla V(\gamma(s)) - \gamma''(s)] ds. \end{aligned}$$

Hence a descent direction is given by:

$$\varphi(s) = -\alpha \gamma^{(4)}(s) - \beta \nabla V(\gamma(s)) + \gamma''(s),$$

which yields the gradient descent equation:

$$\frac{\partial \gamma(s; t)}{\partial t} = -\alpha \gamma^{(4)}(s) - \beta \nabla V(\gamma(s)) + \gamma''(s)$$

and the corresponding iterative scheme:

$$\gamma^{n+1}(s) = \gamma^n(s) + \Delta t [-\alpha \gamma^{n(4)}(s) - \beta \nabla V(\gamma^n(s)) + \gamma^{n''}(s)],$$

where Δt is the time step.

Remark. Note that since D is typically two-dimensional, γ and φ are vector-valued functions. Hence the integrations by parts are written in abuse of notation. If the reader is troubled, it is advised that they consider the integrals on each coordinate separately.

(b)

$$E_{snake,2}^{discrete}(\gamma) = \frac{1}{2} \sum_{i=1}^N \frac{[\gamma(s_{i+1}) - \gamma(s_i)]^2}{\sigma^2} + \frac{\alpha}{2} \sum_{i=1}^N \frac{[\gamma(s_{i+1}) - 2\gamma(s_i) + \gamma(s_{i-1}))]^2}{\sigma^4} + \beta \sum_{i=1}^N V(\gamma(s_i)).$$

(c)

$$\begin{aligned} & \frac{\partial E_{snake,2}(\gamma)}{\partial \gamma(s_i)} \\ &= \frac{\partial}{\partial \gamma(s_i)} \left\{ \frac{1}{2} \sum_{j=1}^N \frac{[\gamma(s_{j+1}) - \gamma(s_j)]^2}{\sigma^2} + \frac{\alpha}{2} \sum_{j=1}^N \frac{[\gamma(s_{j+1}) - 2\gamma(s_j) + \gamma(s_{j-1}))]^2}{\sigma^4} + \beta \sum_{j=1}^N V(\gamma(s_j)) \right\} \\ &= \frac{-\gamma(s_{i-1}) + 2\gamma(s_i) - \gamma(s_{i+1})}{\sigma^2} + \frac{\alpha[\gamma(s_{j-2}) - 4\gamma(s_{j-1}) + 6\gamma(s_j) - 4\gamma(s_{j+1}) + \gamma(s_{j+2})]}{\sigma^4} \\ & \quad + \beta \nabla V(\gamma(s_i)). \end{aligned}$$

Hence guided by the gradient descent equation:

$$\frac{\partial \gamma}{\partial t} = -\nabla E_{snake,2}(\gamma),$$

we iteratively minimize $E_{snake,2}(\gamma)$ by updating γ :

$$\begin{aligned} \frac{\gamma^{n+1}(s_i) - \gamma^n(s_i)}{\Delta t} &= \frac{\gamma^n(s_{i-1}) - 2\gamma^n(s_i) + \gamma^n(s_{i+1})}{\sigma^2} \\ & \quad + \frac{\alpha[-\gamma^n(s_{j-2}) + 4\gamma^n(s_{j-1}) - 6\gamma^n(s_j) + 4\gamma^n(s_{j+1}) - \gamma^n(s_{j+2})]}{\sigma^4} \\ & \quad - \beta \nabla V(\gamma^n(s_i)), \end{aligned}$$

where Δt is the time step.