

MMAT5390: Mathematical Image Processing

Assignment 5 Solutions

1. (a) Please check that $\frac{\partial \vec{f}^T \vec{a}}{\partial \vec{f}} = \vec{a}$, $\frac{\partial \vec{b}^T \vec{f}}{\partial \vec{f}} = \vec{b}$ and $\frac{\partial \vec{f}^T A \vec{f}}{\partial \vec{f}} = (A + A^T) \vec{f}$.

We know the minimizer must satisfy

$$\mathcal{D} = \frac{\partial}{\partial \vec{f}} [\vec{f}^T L_1^T L_1 \vec{f} + \vec{f}^T L_2^T L_2 \vec{f} + \lambda (\vec{g} - D \vec{f})^T (\vec{g} - D \vec{f})] = 0$$

where λ is the Lagrange multiplier. Therefore,

$$\begin{aligned} \mathcal{D} &= 0 \\ \Rightarrow 2(L_1^T L_1 + L_2^T L_2) \vec{f} + \lambda (-D^T \vec{g} - D^T \vec{g} + 2D^T D \vec{f}) &= 0 \\ \Rightarrow (\lambda D^T D + L_1^T L_1 + L_2^T L_2) \vec{f} &= \lambda D^T \vec{g}. \end{aligned}$$

- (b) Since L_1 , L_2 are also block-circulant, they are also diagonalizable by W . Denote $W^{-1} D W$ by Λ_D , $W^{-1} L_1 W$ by Λ_{L_1} and $W^{-1} L_2 W$ by Λ_{L_2} . Then

$$\begin{aligned} \lambda D^T D + L_1^T L_1 + L_2^T L_2 &= \lambda D^* D + L_1^* L_1 + L_2^* L_2 \\ &= \lambda (W \Lambda_D W^{-1})^* (W \Lambda_D W^{-1}) + (W \Lambda_{L_1} W^{-1})^* (W \Lambda_{L_1} W^{-1}) \\ &\quad + (W \Lambda_{L_2} W^{-1})^* (W \Lambda_{L_2} W^{-1}) \\ &= \lambda W \Lambda_D^* \Lambda_D W^{-1} + W \Lambda_{L_1}^* \Lambda_{L_1} W^{-1} + W \Lambda_{L_2}^* \Lambda_{L_2} W^{-1} \end{aligned}$$

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

$$\begin{aligned} (\lambda W \Lambda_D^* \Lambda_D W^{-1} + W \Lambda_{L_1}^* \Lambda_{L_1} W^{-1} + W \Lambda_{L_2}^* \Lambda_{L_2} W^{-1}) \mathcal{S}(f) &= \lambda W \Lambda_D^* W^{-1} \mathcal{S}(g), \\ (\lambda \Lambda_D^* \Lambda_D + \Lambda_{L_1}^* \Lambda_{L_1} + \Lambda_{L_2}^* \Lambda_{L_2}) \mathcal{S}(DFT(f)) &= \lambda \Lambda_D^* \mathcal{S}(DFT(g)) \end{aligned}$$

and thus

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

2. (a) After applying mean filter,

$$\tilde{I} = \frac{1}{9} \begin{pmatrix} 24 & 25 & 24 & 23 \\ 25 & 28 & 25 & 24 \\ 24 & 25 & 28 & 25 \\ 23 & 24 & 25 & 24 \end{pmatrix}$$

- (b) After applying median filter,

$$\tilde{I} = \begin{pmatrix} 3 & 3 & 3 & 3 \\ 3 & 3 & 3 & 3 \\ 3 & 3 & 3 & 3 \\ 3 & 3 & 3 & 3 \end{pmatrix}$$

- (c)

$$\tilde{I} = K * I = \begin{pmatrix} 2 & 0 & -\frac{1}{2} & -2 \\ 0 & 1 & 0 & -\frac{1}{2} \\ -\frac{1}{2} & 0 & 1 & 0 \\ -2 & -\frac{1}{2} & 0 & 2 \end{pmatrix}$$

3. Note that

$$\begin{aligned}
I_{sharp}(0, 0) &= I(0, 0) - \Delta I(0, 0) \\
4 &= 1 - (-4 + a + 0 + b + 0) \\
a + b &= 1 \\
I_{sharp}(0, 2) &= I(0, 2) - \Delta I(0, 2) \\
-2 &= 0 - (0 + 0 + 0 + 0 + b) \\
b &= 2 \\
I_{sharp}(1, 3) &= I(1, 3) - \Delta I(1, 3) \\
1 &= 0 - (0 + c + 0 + a + 0) \\
a + c &= -1
\end{aligned}$$

By solving the above equations, we have $a = -1, b = 2$ and $c = 0$.

4. (a) Suppose f is a minimizer of E . Then for any $v : \Omega \rightarrow \mathbb{R}$,

$$\begin{aligned}
0 &= \left. \frac{\partial}{\partial t} \right|_{t=0} E(f + tv) \\
&= \int_{\Omega} \left. \frac{\partial}{\partial t} \right|_{t=0} [(f + tv - g)^2 + \lambda \|\nabla(f + tv)\|_2^2] dx dy \\
&= \int_{\Omega} [2v(f - g) + \lambda \left. \frac{\partial}{\partial t} \right|_{t=0} \langle \nabla f + t\nabla v, \nabla f + t\nabla v \rangle] dx dy \\
&= \int_{\Omega} [2v(f - g) + \lambda \left. \frac{\partial}{\partial t} \right|_{t=0} (\|\nabla f\|_2^2 + 2t\langle \nabla f, \nabla v \rangle + t^2\|\nabla v\|_2^2)] dx dy \\
&= 2 \int_{\Omega} [v(f - g) + \lambda \langle \nabla f, \nabla v \rangle] dx dy \\
&= 2 \int_{\Omega} v[f - g - \lambda \nabla \cdot (\nabla f)] + 2\lambda \int_{\partial\Omega} v \langle \nabla f, \vec{n} \rangle ds. \quad (*)
\end{aligned}$$

Since v is arbitrarily chosen, f satisfies:

$$\begin{cases} -\lambda \nabla^2 f(x, y) + f(x, y) = g(x, y) & \text{if } (x, y) \in \Omega, \\ \langle \nabla f(x, y), \vec{n}(x, y) \rangle = 0 & \text{if } (x, y) \in \partial\Omega. \end{cases}$$

(b) Suppose f satisfies the above PDE. Then for any $h : \Omega \rightarrow \mathbb{R}$,

$$\begin{aligned}
E(h) - E(f) &= \int_{\Omega} [(h - g)^2 - (f - g)^2 + \lambda \|\nabla h\|_2^2 - \lambda \|\nabla f\|_2^2] dx dy \\
&= \int_{\Omega} [h^2 - 2gh + g^2 - f^2 + 2fg - g^2 + \lambda (\|\nabla h - \nabla f\|_2^2 - 2\|\nabla f\|_2^2 + 2\langle \nabla h, \nabla f \rangle)] dx dy \\
&= \int_{\Omega} [(h - f)^2 + 2hf - 2f^2 - 2gh + 2fg + \lambda (\|\nabla h - \nabla f\|_2^2 + 2\langle \nabla f, \nabla h - \nabla f \rangle)] dx dy \\
&\geq 2 \int_{\Omega} [(f - g)(h - f) + \lambda \langle \nabla f, \nabla h - \nabla f \rangle] dx dy \\
&= 2 \int_{\Omega} [(f - g)(h - f) - \lambda (h - f) \nabla^2 f] dx dy + 2 \int_{\partial\Omega} (h - f) \langle \nabla f, \vec{n} \rangle ds = 0.
\end{aligned}$$

Hence f minimizes E .

(c) From (*), a descent direction of E at f is given by $-v$ where:

$$v(x, y) = \begin{cases} -f(x, y) + g(x, y) + \lambda \nabla^2 f(x, y) & \text{if } (x, y) \in \Omega, \\ -\langle \nabla f(x, y), \vec{n}(x, y) \rangle & \text{if } (x, y) \in \partial\Omega. \end{cases}$$

Hence given $f : \Omega \rightarrow \mathbb{R}$, the gradient descent scheme for minimizing E is given by:

$$\frac{\partial f(x, y; t)}{\partial t} = \begin{cases} -f(x, y) + g(x, y) + \lambda \nabla^2 f(x, y) & \text{if } (x, y) \in \Omega, \\ -\langle \nabla f(x, y), \vec{n}(x, y) \rangle & \text{if } (x, y) \in \partial\Omega. \end{cases}$$

5. (Coding assignment, optional)

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1 if t >= n/(2^p) && t < (n+0.5)/(2^p)
2     H(i, j) = sqrt(2)^p;
3 elseif t >= (n+0.5)/(2^p) && t < (n+1)/(2^p)
4     H(i, j) = -sqrt(2)^p;
5 end

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1 img = img + G(i, j) * Ht(:, i) * H(j, :);

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