

# Image Processing 2, Exercise 2

## 1 Structure tensor

[intermediate] An example of the power of the structure tensor to encode local gradient information about an image.

(a) Consider computing over an image  $f$  the delocalized structure tensor  $\mathbf{J}$  characterized by the constant-valued observation window  $w = 1$ . Use Parseval to give an expression of  $\mathbf{J}$  where  $\hat{f}$  appears instead of  $f$ .

**Solution:**

$$\begin{aligned}
 \mathbf{J} &= \int_{\mathbb{R}^2} \nabla f(\mathbf{x}) (\nabla f(\mathbf{x}))^H d\mathbf{x} \\
 \mathbf{J} &= \int_{\mathbb{R}^2} \begin{pmatrix} \frac{\partial f(\mathbf{x})}{\partial x_1} \left( \frac{\partial f(\mathbf{x})}{\partial x_1} \right)^* & \frac{\partial f(\mathbf{x})}{\partial x_1} \left( \frac{\partial f(\mathbf{x})}{\partial x_2} \right)^* \\ \frac{\partial f(\mathbf{x})}{\partial x_2} \left( \frac{\partial f(\mathbf{x})}{\partial x_1} \right)^* & \frac{\partial f(\mathbf{x})}{\partial x_2} \left( \frac{\partial f(\mathbf{x})}{\partial x_2} \right)^* \end{pmatrix} dx_1 dx_2 \\
 &\stackrel{(1)}{=} \frac{1}{4\pi^2} \int_{\mathbb{R}^2} \begin{pmatrix} (j\omega_1 \hat{f}(\omega)) (j\omega_1 \hat{f}(\omega))^* & (j\omega_1 \hat{f}(\omega)) (j\omega_2 \hat{f}(\omega))^* \\ (j\omega_2 \hat{f}(\omega)) (j\omega_1 \hat{f}(\omega))^* & (j\omega_2 \hat{f}(\omega)) (j\omega_2 \hat{f}(\omega))^* \end{pmatrix} d\omega_1 d\omega_2 \\
 &\stackrel{(2)}{=} \frac{1}{4\pi^2} \int_{\mathbb{R}^2} \begin{pmatrix} \omega_1^2 & \omega_1 \omega_2 \\ \omega_2 \omega_1 & \omega_2^2 \end{pmatrix} |\hat{f}(\omega)|^2 d\omega_1 d\omega_2
 \end{aligned}$$

where (1) uses the differentiation property of Fourier transform; (2) uses the basic properties of complex numbers.

(b) Let an image be  $f(\mathbf{x}) = \text{sinc}(2x_1 + 3x_2) \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x_1^2 + 4x_1 x_2 + 4x_2^2)}$ . Determine its gradient  $\nabla f$ . Hint: To avoid direct calculation of  $\nabla f$  which is complex, notice that image  $f$  can be written as applying an affine transformation  $\mathbf{A}$  to another image  $g$  that has a much simpler form, try to find out  $\mathbf{A}$  and  $g$ .

**Solution:** Observe that  $f(\mathbf{x})$  can be written as

$$f(\mathbf{x}) = \text{sinc}(2x_1 + 3x_2) \frac{1}{\sqrt{2\pi}} e^{-\frac{(x_1+2x_2)^2}{2}}.$$

Let  $y_1 = 2x_1 + 3x_2$ ,  $y_2 = x_1 + 2x_2$ , and  $g(y_1, y_2) = \text{sinc}(y_1) \frac{1}{\sqrt{2\pi}} e^{-\frac{y_2^2}{2}}$ , then  $f(\mathbf{x}) = g(\mathbf{y}) = g(\mathbf{A}\mathbf{x})$ , where  $\mathbf{A} = \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix}$ .

Hence,

$$\begin{aligned}
 \nabla f(\mathbf{x}) &= \nabla g(\mathbf{A}\mathbf{x}) \\
 &= (\mathbf{A}\mathbf{x})'^T g'(\mathbf{A}\mathbf{x}) \\
 &= \mathbf{A}^T g'(\mathbf{A}\mathbf{x}) \\
 &= \mathbf{A}^T g'(\mathbf{y})|_{\mathbf{y}=\mathbf{A}\mathbf{x}}.
 \end{aligned}$$

The gradient of  $g$  is

$$\begin{aligned}
 g'(\mathbf{y}) &= \nabla \left( \text{sinc}(y_1) \frac{1}{\sqrt{2\pi}} e^{-\frac{y_2^2}{2}} \right) \\
 &= \frac{1}{\sqrt{2\pi}} \begin{pmatrix} \frac{\cos(\pi y_1) \pi \pi y_1 - \sin(\pi y_1) \pi}{(\pi y_1)^2} \\ \text{sinc}(y_1) (-y_2) \end{pmatrix} e^{-\frac{y_2^2}{2}}.
 \end{aligned}$$

Combining all the information above, we obtain

$$\begin{aligned}
 \nabla f(\mathbf{x}) &= \mathbf{A}^\top g'(\mathbf{y})|_{\mathbf{y}=\mathbf{A}\mathbf{x}} \\
 &= \frac{1}{\sqrt{2\pi}} \left( \begin{array}{l} 2 \frac{\cos(\pi(2x_1+3x_2)) - \text{sinc}(2x_1+3x_2)}{2x_1+3x_2} - \text{sinc}(2x_1+3x_2)(x_1+2x_2) \\ 3 \frac{\cos(\pi(2x_1+3x_2)) - \text{sinc}(2x_1+3x_2)}{2x_1+3x_2} - 2\text{sinc}(2x_1+3x_2)(x_1+2x_2) \end{array} \right) \\
 &\times e^{-\frac{(x_1+2x_2)^2}{2}}.
 \end{aligned}$$

(c) Determine the Fourier transform  $\hat{f}$  of the image  $f$ .

**Solution:**

$$\begin{aligned}
 \hat{g}(\boldsymbol{\omega}) &= \mathcal{F}_{\mathbf{x}}\{\text{sinc}(x_1) \frac{1}{\sqrt{2\pi}} e^{-\frac{x_1^2}{2}}\}(\boldsymbol{\omega}) \\
 &= \mathcal{F}_{x_1}\{\text{sinc}(x_1)\}(\omega_1) \mathcal{F}_{x_2}\{\frac{1}{\sqrt{2\pi}} e^{-\frac{x_2^2}{2}}\}(\omega_2) \\
 &= \text{rect}\left(\frac{\omega_1}{2\pi}\right) e^{-\frac{\omega_2^2}{2}}
 \end{aligned}$$

$$\begin{aligned}
 \hat{f}(\boldsymbol{\omega}) &\stackrel{(1)}{=} |\det(\mathbf{A})|^{-1} \hat{g}(\mathbf{A}^{-\top} \boldsymbol{\omega}) \\
 &= \text{rect}\left(\frac{2\omega_1 - \omega_2}{2\pi}\right) e^{-\frac{1}{2}(9\omega_1^2 - 12\omega_1\omega_2 + 4\omega_2^2)}
 \end{aligned}$$

where (1) uses the affine transformation property of Fourier transform in the appendix.

(d) Determine the value of the delocalized structure tensor associated to  $f$ . You may want to take advantage of  $\int_{\mathbb{R}} e^{-x^2} dx = \sqrt{\pi}$  and  $\int_{\mathbb{R}} x^2 e^{-x^2} dx = \frac{1}{2} \sqrt{\pi}$ .

**Solution:** Let

$$\boldsymbol{\varphi} = \mathbf{A}^{-\top} \boldsymbol{\omega} \quad \boldsymbol{\omega} = \mathbf{A}^\top \boldsymbol{\varphi}$$

Combining the result of part (a) and (c),

$$\begin{aligned}
 \mathbf{J} &= \frac{1}{4\pi^2} \int_{\mathbb{R}^2} \left( \begin{array}{cc} \omega_1^2 & \omega_1\omega_2 \\ \omega_1\omega_2 & \omega_2^2 \end{array} \right) \left| \text{rect}\left(\frac{2\omega_1 - \omega_2}{2\pi}\right) e^{-\frac{1}{2}(9\omega_1^2 - 12\omega_1\omega_2 + 4\omega_2^2)} \right|^2 d\omega_1 d\omega_2 \\
 &\stackrel{(1)}{=} \frac{1}{4\pi^2} \int_{\mathbb{R}^2} \left( \begin{array}{cc} (2\varphi_1 + \varphi_2)^2 & (2\varphi_1 + \varphi_2)(3\varphi_1 + 2\varphi_2) \\ (3\varphi_1 + 2\varphi_2)(2\varphi_1 + \varphi_2) & (3\varphi_1 + 2\varphi_2)^2 \end{array} \right) \text{rect}\left(\frac{\varphi_1}{2\pi}\right) e^{-\varphi_2^2} d\varphi_1 d\varphi_2 \\
 &\stackrel{(2)}{=} \frac{1}{4\pi^2} \int_{-\infty}^{\infty} \int_{-\pi}^{\pi} \left( \begin{array}{cc} 4\varphi_1^2 + 4\varphi_1\varphi_2 + \varphi_2^2 & 6\varphi_1^2 + 7\varphi_1\varphi_2 + 2\varphi_2^2 \\ 6\varphi_1^2 + 7\varphi_1\varphi_2 + 2\varphi_2^2 & 9\varphi_1^2 + 12\varphi_1\varphi_2 + 4\varphi_2^2 \end{array} \right) d\varphi_1 e^{-\varphi_2^2} d\varphi_2 \\
 &= \frac{1}{4\pi^2} \int_{-\infty}^{\infty} \left( \begin{array}{cc} \frac{8}{3}\pi^3 + 2\pi\varphi_2^2 & 4\pi^3 + 4\pi\varphi_2^2 \\ 4\pi^3 + 4\pi\varphi_2^2 & 6\pi^3 + 8\pi\varphi_2^2 \end{array} \right) e^{-\varphi_2^2} d\varphi_2 \\
 &= \frac{1}{12\sqrt{\pi}} \left( \begin{array}{cc} 3 + 8\pi^2 & 6 + 12\pi^2 \\ 6 + 12\pi^2 & 12 + 18\pi^2 \end{array} \right).
 \end{aligned}$$

(1) applies the change of variables,  $d\omega_1 d\omega_2 = |\det(\mathbf{A}^\top)| d\varphi_1 d\varphi_2$ . (2) uses the property that function  $\text{rect}(\frac{\varphi_1}{2\pi})$  is supported on  $[-\pi, \pi]$ .

(e) Give the delocalized gradient energy of  $f$ .

**Solution:**

$$\begin{aligned}
 E &= \text{tr}(\mathbf{J}) \\
 &= [\mathbf{J}]_{2,2} + [\mathbf{J}]_{1,1} \\
 &= \frac{15 + 26\pi^2}{12\sqrt{\pi}}
 \end{aligned}$$

(f) Give the delocalized coherency of  $f$ .

**Solution:**

$$\begin{aligned}
 C &= \frac{1}{E} \sqrt{([\mathbf{J}]_{2,2} - [\mathbf{J}]_{1,1})^2 + 4 [\mathbf{J}]_{1,2}^2} \\
 &= \frac{\sqrt{225 + 756\pi^2 + 676\pi^4}}{15 + 26\pi^2}
 \end{aligned}$$

## 2 Spline Interpolation 1D

*[basic] Interpolating the samples of a function using B-splines. Interpolation is fundamental in image processing, because we often want to move back and forth between continuous-domain signals and their discrete-domain representations.*

Assume that  $f(x) = (10 - |6x + 3|) \text{rect}(\frac{x}{4} + \frac{1}{8})$ . Find the quadratic spline coefficients  $\{c[m]\}_{m \in \mathbb{Z}}$  such that  $s(x) = \sum_{m \in \mathbb{Z}} c[m] \beta^2(x - m)$  satisfies the interpolation condition  $s[k] = f[k]$  for all  $k \in \mathbb{Z}$ .

**Solution:** Let  $F(z)$  be the z-transform of the sequence  $f[k]$  for  $k \in \mathbb{Z}$ . It is given as

$$F(z) = z^2 + 7z + 7 + z^{-1},$$

because  $f[k]$  is nonzero only at  $k = 0 \Rightarrow f[0] = 7$ ,  $k = 1 \Rightarrow f[1] = 1$ ,  $k = -1 \Rightarrow f[-1] = 7$ ,  $k = -2 \Rightarrow f[-2] = 1$ .

We now want that  $s[k] = f[k]$  for  $k \in \mathbb{Z}$ . This means their z-transform are equal too *i.e.*,

$$\forall k \in \mathbb{Z} \quad f[k] = s[k] \implies F(z) = S(z).$$

The z-transform of  $s[k]$  is given by

$$\begin{aligned}
 S(z) &= C(z)B(z) \\
 S(z) &= C(z) \frac{z + 6 + z^{-1}}{8}.
 \end{aligned}$$

From the previous equalities we get,

$$\begin{aligned}
 C(z) &= \frac{8S(z)}{z + 6 + z^{-1}}, \\
 &= \frac{8F(z)}{z + 6 + z^{-1}}, \\
 &= 8 \frac{z(z + 6 + z^{-1}) + (z + 6 + z^{-1})}{z + 6 + z^{-1}}, \\
 &= 8z + 8.
 \end{aligned}$$

This gives

$$c[m] = 8\delta[m] + 8\delta[m + 1].$$

This example shows that given a sample of a function (in this case of  $f(x)$ ), we can create a new function  $s(x)$  using splines such that its samples are the same as of  $f(x)$ .

**Alternative solution:**

Here, we present an alternative solution based on the implementation presented in slide 7-24 and the table of transform function of B-splines in slide 7-34. Because we chose quadratic splines,  $n = 2$  and  $a = -3 + 2\sqrt{2}$ .

We consider the discrete signal

$$f[k] = \left[ \dots \ 0 \ 0 \ 1 \ 7 \ \boxed{7} \ 1 \ 0 \ 0 \ \dots \right] \quad (1)$$

and

$$f[k] = b * c[k],$$

$$\text{where } b[k] = \left[ \dots \ \frac{1}{8} \ \boxed{\frac{6}{8}} \ \frac{1}{8} \dots \right].$$

Let  $y_1$  be the result of the causal filter on  $f$ , let  $y_2$  be the result of the anti-causal filter on  $y_1$ .

Be very careful! They are the results of the filtering, not the filters themselves! Using the cascade of a causal and anti-causal filter, we obtain the following partial signals.

Causal filter (recall  $y_1[k] = f[k] + ay_1[k - 1]$  from the causal filter cascade)

$$\begin{aligned} y_1[k] &= 0, \quad k \leq -3 \text{ since filter is causal and } f[k] = 0 \text{ for } k \leq -3 \\ y_1[-2] &= f[-2] + z_1 y_1[-3] \\ &= 1 \\ y_1[-1] &= f[-1] + a y_1[-2] \\ &= 7 - 3 + 2\sqrt{2} \\ &= 4 + 2\sqrt{2} \\ y_1[0] &= f[0] + a y_1[-1] \\ &= 3 + 2\sqrt{2} \\ y_1[1] &= f[1] + a y_1[0] \\ &= 0 \\ y_1[k] &= 0 \quad \text{for } k > 1 \end{aligned}$$

and we have

$$y_1[k] = \left[ \dots \ 0 \ 1 \ 4 + 2\sqrt{2} \ \boxed{3 + 2\sqrt{2}} \ 0 \ \dots \right].$$

Anti-causal filter (recall  $y_2[k] = y_1[k] + ay_2[k + 1]$  from the filter cascade)

$$\begin{aligned}
y_2[k] &= 0 \quad k \geq 1 \text{ since filter is anti-causal and } y_1[k] = 0 \text{ for } k \geq 1 \\
y_2[0] &= y_1[0] + ay_2[1] \\
&= 3 + 2\sqrt{2} \\
y_2[-1] &= y_2[-1] + ay_2[0] \\
&= 4 + 2\sqrt{2} + (-3 + 2\sqrt{2})(3 + 2\sqrt{2}) \\
&= 4 + 2\sqrt{2} + 8 - 9 \\
&= 3 + 2\sqrt{2} \\
y_2[-2] &= y_1[-2] + ay_2[-1] \\
&= 1 + (-3 + 2\sqrt{2})(3 + 2\sqrt{2}) \\
&= 1 - 1 \\
&= 0 \\
y_2[k] &= 0 \quad \text{for } k < -2
\end{aligned}$$

and we have

$$y_2[k] = \left[ \cdots 0 \quad 0 \quad 3 + 2\sqrt{2} \quad \boxed{3 + 2\sqrt{2}} \quad 0 \quad \cdots \right].$$

Gain ( $G = \frac{8}{3+2\sqrt{2}}$ )

$$c[k] = \frac{8}{3+2\sqrt{2}} y_2[k] = \left[ \cdots 0 \quad 0 \quad 8 \quad \boxed{8} \quad 0 \quad \cdots \right]$$

We can finally verify that

$$f[k] = b * c[k]$$

by computing the FIR convolution:

$$\begin{array}{c|cccccccc}
& 0 & 0 & 0 & 8 & \boxed{8} & 0 & 0 & 0 \\
\hline
\frac{1}{8} & 1 & \boxed{6} & 1 & 1 & \boxed{6} & 1 & 1 & 1 \\
\frac{1}{8} & & 1 & \boxed{6} & 1 & 1 & \boxed{6} & 1 & 1 \\
\frac{1}{8} & & & 1 & \boxed{6} & 1 & 1 & \boxed{6} & 1 \\
\frac{1}{8} & & & & 1 & \boxed{6} & 1 & 1 & \boxed{6} \\
\frac{1}{8} & & & & & 1 & \boxed{6} & 1 & 1 \\
\frac{1}{8} & & & & & & 1 & \boxed{6} & 1 \\
\hline
& 0 & 0 & 1 & 7 & \boxed{7} & 1 & 0 & 0
\end{array}$$