# CV192, HW # 4

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# 1 Theoretical Questions

#### 1.1

We will show the that convolution is associative:

$$\left(a*b\right)*c_{\left(i,j\right)} = \sum_{k,l} \left(a*b\right)\left(i-k,j-l\right)c\left(k,l\right) = \sum_{k,l} \sum_{k',l'} a\left(i-k-k',j-l-l'\right)b\left(k',l'\right)c\left(k,l\right)$$

We define  $k^* = k + k'$  and  $l^* = l + l'$  and get:

$$\sum_{k^*,l^*} \sum_{k,l} a\left(i-k^*,j-l^*\right) b\left(k^*-k,l^*-l\right) c\left(k,l\right) = \sum_{k^*,l^*} a\left(i-k^*,j-l^*\right) \left(b*c\right) (k^*,l^*) = a*(b*c)_{(i,j)}$$

# 1.2

We will show that  $h * \delta = \delta * h = h$ .

$$h * \delta = \sum_{k,l} h(i - k, j - l)\delta(k, l) = h(i - 0, j - 0) \cdot 1 = h(i, j)$$
$$\delta * h = \sum_{k,l} \delta(i - k, j - l)h(k, l) = h(k, l) \stackrel{*}{=} h(i, j)$$

\* For i=k and j=l.

# 1.3

At first we will define  $h_{flipped}$  for the convenience of performing convolution:

$$\mathbf{h}_{flipped} = \begin{bmatrix} h_{1,1} & h_{1,2} & h_{1,3} \\ h_{2,1} & h_{2,2} & h_{2,3} \\ h_{3,1} & h_{3,2} & h_{3,3} \end{bmatrix}$$

The values of H are determined by h, we define the relative position as follows :

$$H_{i,j} = \begin{cases} h_{1,1}, & j = i-1-N \wedge i mod N - 1 \geq 0 \\ h_{1,2}, & j = i-N \\ h_{1,3}, & j = i+1-N \wedge i mod N + 1 < N \\ h_{2,1}, & j = i-1 \wedge i mod N - 1 \geq 0 \\ h_{2,2}, & j = i \\ h_{2,3}, & j = i+1 \wedge i mod N + 1 < N \\ h_{3,1}, & j = i-1+N \wedge i mod N - 1 \geq 0 \\ h_{3,2}, & j = i+N \\ h_{3,3}, & j = i+1+N \wedge i mod N + 1 < N \\ 0, & otherwise \end{cases}$$

#### 1.4

i) In order to recover x, we need H to be invertible (not singular), so we get:

$$y = Hx \Longrightarrow x = H^{-1}y$$

ii) According to the given data we know that:

$$\mathbf{h} = \mathbf{h}_{flipped} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Hence, the matrix H we constructed in the previous exercise will be the Identity matrix  $I = H \in M_{MN \times MN}$  so we get that H is invertible as wanted. The recover of x will be:

$$x = H^{-1}y \Rightarrow x = Iy \Rightarrow x = y$$

#### 1.5

The effect of convolving with:

is moving each pixel 2 pixels to the right and 1 pixel down, we can inspect this by the 1 location in the filter (2 positions to the right and 1 position down from the center) or from the convolved image:

#### 1.6

i) Let Y be a square orthogonal matrix, we can say that:

$$det(Y) = \pm 1.$$

Proof:

$$\begin{split} YY^T &= I \\ det(I) &= 1 \\ det(YY^T) &= det(I) = 1 \\ det(YY^T) &= det(Y)det(Y^T) = 1 \Rightarrow det(Y) = det(Y^T) \\ det(Y) &= \pm 1 \end{split}$$

ii) In order to prove that  $Y^T$  is orthogonal we need to show that  $Y^T(Y^T)^T = I$ :

$$YY^T = I \Rightarrow Y^T = Y^{-1}$$

$$Y^{T}(Y^{T})^{T} = Y^{T}Y = Y^{-1}Y = I$$

iii)  $Y, Y^T$  are square orthogonal matrices,  $y_i^T y_i$  is the i,i entry in  $Y^T Y$ , therefore:

$$Y^TY = I \Rightarrow y_i^T y_i = 1$$

iv) 
$$\| \boldsymbol{y}_i \|_{\ell_2} = \sqrt{\sum_{j=1}^n |y_{j,i}|^2} = \sqrt{y_i^T y_i} = \sqrt{1} = 1$$

- v)  $y_i^T y_j$  is equal to the (i,j) entry in  $Y^T Y = I$ , therefore for  $i \neq j$ ,  $y_i^T y_j = 0$ . Meaning that the vectors  $y_i^T$  and  $y_j$  are orthogonal to each other and the angle between them is 90.
- vi) Let G be the group for  $n \times n$  orthogonal matrices, we will prove the 3 qualities of such group:
- 1)  $I \in G$

$$II^T = I \Rightarrow I \in G$$

 $2)\ A\in G, B\in G\Rightarrow AB\in G$ 

$$A, B \in G \Rightarrow (AB)(AB)^T = ABB^TA^T = AIA^T = AA^T = I \Rightarrow AB \in G$$

3)  $A \in G$ ,  $A^{-1}$  exists and  $A^{-1} \in G$ 

$$A \in G \Rightarrow AA^T = I \Rightarrow A^{-1} = A^T$$

 $A^T$  is orthogonal so:

$$A^T \in G \Rightarrow A^T = A^{-1} \in G$$

# 1.7

Let K be an  $n \times n$  separable filter. We will show that K is not invertible:

$$det(K) = det(USV^T) = det(U)det(S)det(V^T) = det(U) \cdot 0 \cdot det(V^T) = 0$$

A matrix with determinate equal to zero is not invertiable.

#### 1.8

Bilateral filtering is not a linear operation.

A linear operation is one that holds the following constraint:

$$h(f + \lambda q) = h(f) + \lambda h(q)$$

If we look at the definition of the bilateral filter:

$$g(x) = \frac{1}{c} \sum_{x_i} f^r(|I(x) - I(x_i)|) f^d(|x - x_i|) I(x_i)$$

we can see that  $f^r$ ,  $f^d$  may be defined as Gaussian, hence the functions will contain a multiplication of exponent function with squared error difference, which is obviously not linear. If we take the definition of linear operator over filters we get that the following must hold:

$$(c_1I_1 + c_2I_2) * h = c_1(i_1 * h) + c_2(I_2 * h)$$

While on the left side of the equation that exponent power will be:

$$(c_1I_1(x) + c_2I_2(x) - c_1I_1(x_i) + c_2I_2(x_i))^2$$

And on the right side we will have:

$$(c_1I_1(x) - c_1I_1(x_i)^2 + (c_2I_2(x) - c_2I_2(x_i))^2$$

That you can easily see is not equal.

# 1.9

We will calculate the Laplacian of an isotropic Gaussian:

$$G(x, y, \sigma) = \frac{1}{2\pi\sigma^2} exp(-\frac{x^2 + y^2}{2\sigma^2})$$

At, first we will calculate the first partial derivatives:

$$\frac{\partial G(x,y,\sigma)}{\partial x} = \frac{1}{2\pi\sigma^2} \cdot \left(-\frac{x}{\sigma^2}\right) \cdot exp\left(-\frac{x^2 + y^2}{2\sigma^2}\right)$$

$$\frac{\partial G(x,y,\sigma)}{\partial y} = \frac{1}{2\pi\sigma^2} \cdot (-\frac{y}{\sigma^2}) \cdot exp(-\frac{x^2+y^2}{2\sigma^2})$$

The second derivatives:

$$\begin{split} \frac{\partial^2 G(x,y,\sigma)}{\partial x} &= \frac{1}{2\pi\sigma^2} \cdot (-\frac{x^2}{\sigma^2}) \cdot exp(-\frac{x^2+y^2}{2\sigma^2}) + \frac{1}{2\pi\sigma^2} \cdot (-\frac{x}{\sigma^2}) \cdot (exp(-\frac{x^2+y^2}{2\sigma^2}) \cdot (-\frac{x}{\sigma^2})) \\ &= \frac{1}{2\pi\sigma^2} \cdot (\frac{x^2}{\sigma^4} - \frac{1}{\sigma^2}) \cdot exp(-\frac{x^2+y^2}{2\sigma^2}) \end{split}$$

$$\begin{split} \frac{\partial^2 G(x,y,\sigma)}{\partial y} &= \frac{1}{2\pi\sigma^2} \cdot (-\frac{x^2+y^2}{\sigma^2}) + \frac{1}{2\pi\sigma^2} \cdot (-\frac{y}{\sigma^2}) \cdot (exp(-\frac{x^2+y^2}{2\sigma^2}) \cdot (-\frac{y}{\sigma^2})) \\ &= \frac{1}{2\pi\sigma^2} \cdot (\frac{y^2}{\sigma^4} - \frac{1}{\sigma^2}) \cdot exp(-\frac{x^2+y^2}{2\sigma^2}) \end{split}$$

Finally we get:

$$\begin{split} \nabla^2 G(x,y,\sigma) &= \frac{\partial^2 G(x,y,\sigma)}{\partial x} + \frac{\partial^2 G(x,y,\sigma)}{\partial y} \\ &= \frac{1}{2\pi\sigma^2} \cdot (\frac{x^2+y^2}{\sigma^4} - \frac{2}{\sigma^2}) \cdot exp(-\frac{x^2+y^2}{2\sigma^2}) = (\frac{x^2+y^2}{\sigma^4} - \frac{2}{\sigma^2}) \cdot G(x,y,\sigma) \end{split}$$

As requested.

# 1.10

Given that:

$$x' = x\cos\theta - y\sin\theta$$
,

$$y' = xsin\theta + ycos\theta$$

We will show that:

$$\nabla^2 I(x,y) = \nabla^2 I(x'(x,y),y'(x.y))$$

Or equivalently:

$$\frac{\partial^2 I}{\partial x^2} + \frac{\partial^2 I}{\partial y^2} = \frac{\partial^2 I}{\partial x'^2} + \frac{\partial^2 I}{\partial y'^2}$$

Proof:

$$\begin{split} \frac{\partial^2 I}{\partial x^2} &= \frac{\partial}{\partial x} \frac{\partial I}{\partial x} \\ &= \frac{\partial}{\partial x} (\frac{\partial I}{\partial x'} \frac{\partial x'}{\partial x} + \frac{\partial I}{\partial y'} \frac{\partial y'}{\partial x}) \\ &= \frac{\partial}{\partial x} (\frac{\partial I}{\partial x'} \cos\theta + \frac{\partial I}{\partial y'} \sin\theta) \\ &= \frac{\partial I}{\partial x'} \frac{\partial}{\partial x} \cos\theta + \frac{\partial I}{\partial y'} \frac{\partial}{\partial x} \sin\theta \\ &= \frac{\partial I}{\partial x'} \cos\theta (\frac{\partial I}{\partial x'} \cos\theta + \frac{\partial I}{\partial y'} \sin\theta) + \frac{\partial I}{\partial y'} \sin\theta (\frac{\partial I}{\partial x'} \cos\theta + \frac{\partial I}{\partial y'} \sin\theta) \\ &= \frac{\partial^2 I}{\partial x'^2} \cos^2\theta + 2(\frac{\partial^2 I}{\partial x'\partial y'} \sin\theta \cos\theta) + \frac{\partial^2 I}{\partial y'^2} \sin^2\theta \end{split}$$

$$\begin{split} \frac{\partial^2 I}{\partial y^2} &= \frac{\partial}{\partial y} \frac{\partial I}{\partial y} \\ &= \frac{\partial}{\partial y} (\frac{\partial I}{\partial x'} \frac{\partial x'}{\partial y} + \frac{\partial I}{\partial y'} \frac{\partial y'}{\partial y}) \\ &= \frac{\partial}{\partial y} (\frac{\partial I}{\partial x'} (-sin\theta) + \frac{\partial I}{\partial y'} cos\theta) \\ &= \frac{\partial I}{\partial x'} \frac{\partial}{\partial y} (-sin\theta) + \frac{\partial I}{\partial y'} \frac{\partial}{\partial y} cos\theta \\ &= \frac{\partial I}{\partial x'} (-sin\theta) (\frac{\partial I}{\partial x'} (-sin\theta) + \frac{\partial I}{\partial y'} cos\theta) + \frac{\partial I}{\partial y'} cos\theta (\frac{\partial I}{\partial x'} (-sin\theta) + \frac{\partial I}{\partial y'} cos\theta) \\ &= \frac{\partial^2 I}{\partial y'^2} sin^2 \theta - 2 (\frac{\partial^2 I}{\partial x' \partial y'} sin\theta cos\theta) + \frac{\partial^2 I}{\partial y'^2} cos^2 \theta \end{split}$$

Combining our solutions we get:

$$\begin{split} \frac{\partial^2 I}{\partial x^2} + \frac{\partial^2 I}{\partial y^2} &= \frac{\partial^2 I}{\partial x'^2} cos^2 \theta + 2 (\frac{\partial^2 I}{\partial x' \partial y'} sin\theta cos\theta) + \frac{\partial^2 I}{\partial y'^2} sin^2 \theta + \frac{\partial^2 I}{\partial y'^2} sin^2 \theta - 2 (\frac{\partial^2 I}{\partial x' \partial y'} sin\theta cos\theta) + \frac{\partial^2 I}{\partial y'^2} cos^2 \theta \\ &= (cos^2 \theta + sin^2 \theta) (\frac{\partial^2 I}{\partial x'^2} + \frac{\partial^2 I}{\partial y'^2}) = \frac{\partial^2 I}{\partial x'^2} + \frac{\partial^2 I}{\partial y'^2} \end{split}$$

# 1.11

We will show that:

$$GL(n) = \{ \boldsymbol{Q} | \boldsymbol{Q} \in \mathbb{R}^{n \times n}, \det \boldsymbol{Q} \neq 0 \}$$

is a matrix group, In order to do so we will prove the required conditions: G1:

$$det(I_{n\times n}) = 1 \neq 0, I \in \mathbb{R}^{n\times n} \Rightarrow, I \in GL(n)$$

G2: Let  $A, B \in GL(n)$ 

$$\Rightarrow det(A) \neq 0 \neq det(B) \Rightarrow det(AB) = det(A)det(B) \neq 0$$

In addition  $AB \in \mathbb{R}^{n \times n} \Rightarrow AB \in GL(n)$ 

G3: Let  $A \in GL(n)$ 

$$det(A) \neq 0 \Rightarrow A^{-1}exist$$

Since  $A^{-1} \in \mathbb{R}^{n \times n}$  and  $det(A^{-1}) \neq 0, A^{-1} \in GL(n)$ .

# 1.12

We will show that  $GL_+(n) = \{Q | Q \in \mathbb{R}^{n \times n}, \det Q > 0\}$  is a matrix group. G1:

$$I_{n\times n} \in \mathbb{R}^{n\times n}, det(I_{n\times n}) = 1 \Rightarrow I_{n\times n} \in GL_{+}(n)$$

G2: Let  $A,B \in GL_+(n)$ 

$$\Rightarrow A, B \in \mathbb{R}^{n \times n} \Rightarrow AB \in \mathbb{R}^{n \times n}$$

and

$$det(A) > 0, det(B) > 0 \Rightarrow det(AB) = det(A)det(B) > 0 \Rightarrow AB \in GL_{+}(n)$$

G3: Let  $A \in GL_+(n)$ , det(A) > 0 so  $A^{-1}$  exist.

$$A^{-1} \in \mathbb{R}^{n \times n}, \det (A^{-1}) = \frac{1}{\det(A)} > 0 \Rightarrow A^{-1} \in GL_+$$

# 1.13

We will show that  $G = \{Q | Q \in \mathbb{R}^{n \times n}, \det Q < 0\}$  is not a matrix group.

Let  $A, B \in G$ :

$$det(A) < 0, det(B) < 0 \Rightarrow det(AB) = det(A)det(B) > 0 \Rightarrow AB \not\in G$$

Since the G2 condition doesn't hold, G is not a matrix group.

#### 1.14

Let  $A, B \in R_{n \times n}$  symmetric matrices. We will show that usually  $AB \neq (AB)^T$ 

$$(AB)^T = B^T A^T = BA$$

Hence, AB has to be symmetric as well, which is usually not true.

# 1.15

$$US(n) = \{Q|Q = SI_{n \times n}, S \in \mathbb{R}_{>0}\}$$

We will prove that US is a matrix group:

G1: We choose  $S = 1 \in \mathbb{R}_{>0}$ , therefore we get:

$$Q = SI_{n \times n} = I_{n \times n}, det(Q) = det(I_{n \times n}) = 1 \Rightarrow I_{n \times n} \in US(n)$$

G2: Let  $A, B \in US(n)$  Hence the definition of A, B will be:

$$A = S_1 I_{n \times n}, S1 \in \mathbb{R}_{>0}$$

$$B = S_2 I_{n \times n}, S2 \in \mathbb{R}_{>0}$$

Therefore the definition of AB will be:

$$AB = S_1 I_{n \times n} S_2 I_{n \times n} = S_1 S_2 I_{n \times n} I_{n \times n} = S_1 S_2 I_{n \times n}$$

Where  $S_1S_2 \in \mathbb{R}_{>0} \Rightarrow AB \in US(n)$ 

G3: Let  $A \in US(n)$ , therefore:

$$A = SI_{n \times n}, S \in \mathbb{R}_{>0}$$

It is easy to see that  $det(A) = det(SI) = S^n \neq 0$ , hence  $A^{-1}$  exist. Moreover:

$$A^{-1} = (SI_{n \times n})^{-1} = \frac{1}{S}(I_{n \times n})^{-1} = \frac{1}{S}I_{n \times n}$$

Since  $S \in \mathbb{R}_{>0} \Rightarrow \frac{1}{S} \in \mathbb{R}_{>0} \Rightarrow A^{-1} \in US(n)$ .