

5554 - Fall2022 - HW2

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$$g = \begin{bmatrix} -1 & -4 & 7 \\ -2 & 5 & 8 \\ 0 & 6 & 0 \end{bmatrix} \quad h = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}$$

We consider g and h to be surrounded by zeros in all directions.

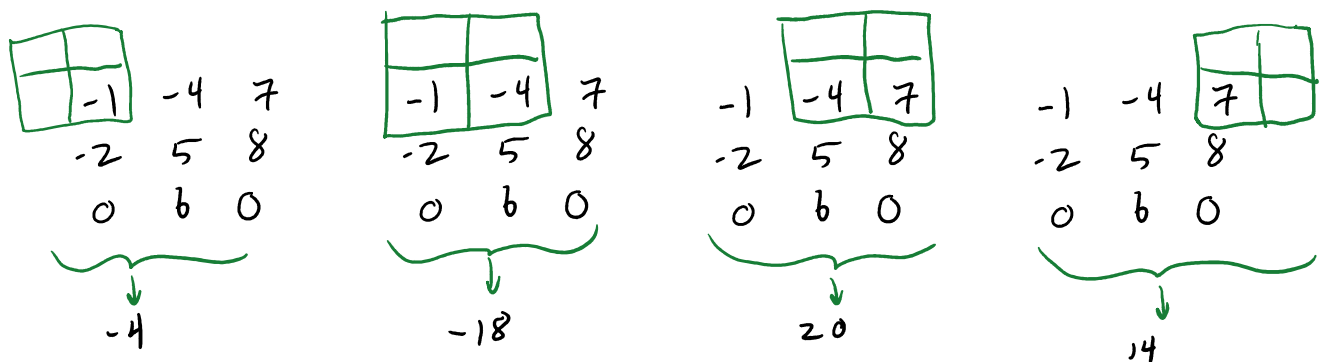
We can choose any convenient coordinate reference frames, as long as we are consistent.

$$a) (g \otimes h)(i,j) = \sum_u \sum_v g(u,v) h(i+u, j+v)$$

We could get the answer directly from the equation,

but it can also be visualized by "sliding" h across g and computing sums of products.

Nonzero values can occur only when both kernels overlap.



Continuing in this way, the answer is

$$g \otimes h = \begin{bmatrix} -4 & -18 & 20 & 14 \\ -11 & 3 & 59 & 23 \\ -6 & 37 & 41 & 8 \\ 0 & 18 & 6 & 0 \end{bmatrix}$$

b)

Similarly, we can "slide" g across h and compute sums of products.

$$h \otimes g = \begin{bmatrix} 0 & 6 & 18 & 0 \\ 8 & 41 & 37 & -6 \\ 23 & 59 & 3 & -11 \\ 14 & 20 & -18 & -4 \end{bmatrix}$$

(Notice that $g \otimes h \neq h \otimes g$)

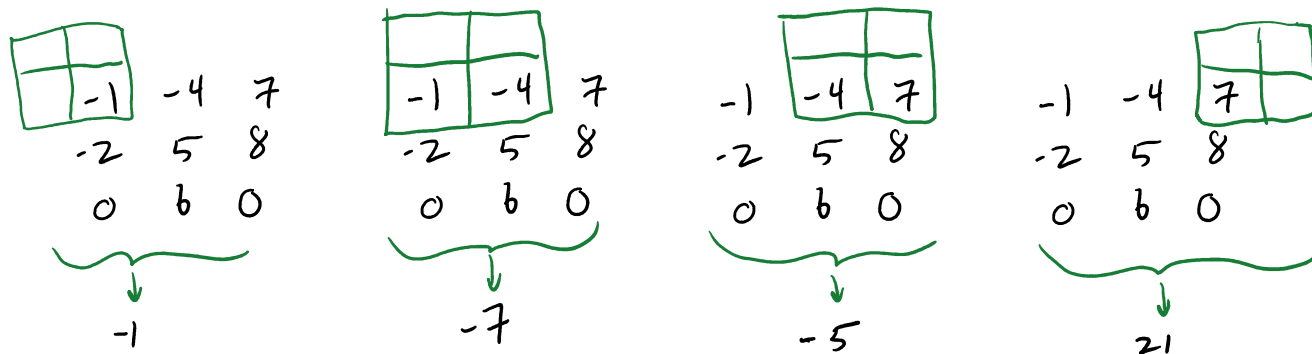
Problem 1 (continued)

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$$c) (g * h)(i, j) = \sum_u \sum_v g(u, v) h(i-u, j-v)$$

Once again, we could get the answer directly from the equation, but it can also be visualized by "sliding" a reversed version of kernel h across g and computing sums of products. Nonzero values can occur only when both kernels overlap.

$$h(-u, -v) = \begin{bmatrix} 4 & 2 \\ 3 & 1 \end{bmatrix}$$



Continuing in this way, the answer is

$$g * h = \begin{bmatrix} -1 & -7 & -5 & 21 \\ -4 & -13 & 21 & 52 \\ -4 & 8 & 54 & 32 \\ 0 & 12 & 24 & 0 \end{bmatrix}$$

(Not required for this problem, but you could also use this same approach to demonstrate that $g * h = h * g$.)

Problem 2

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a) Let $h_1(y) = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$ and $h_2(x) = [-1 \ 0 \ 1]$.

$$\text{Then } h(x,y) = h_1(y) * h_2(x) = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix},$$

which is one of the Sobel filters,

approximating $\frac{\partial}{\partial x}$.

b) We can choose $g_1(x) = \frac{1}{14} [1 \ 3 \ 6 \ 3 \ 1]$

and $g_2(y) = \frac{1}{14} \begin{bmatrix} 1 \\ 3 \\ 6 \\ 3 \\ 1 \end{bmatrix}$. Then a discrete approximation

to a 2D Gaussian filter is given by

$$g(x,y) = g_1(x) * g_2(y) = \frac{1}{196} \cdot \begin{bmatrix} 1 & 3 & 6 & 3 & 1 \\ 3 & 9 & 18 & 9 & 3 \\ 6 & 18 & 36 & 18 & 6 \\ 3 & 9 & 18 & 9 & 3 \\ 1 & 3 & 6 & 3 & 1 \end{bmatrix}$$

Problem 3

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

a) Suppose we convolve $f * G$, for some arbitrary 2D function f .

$$f(x, y) * G(x, y)$$

$$= \iint f(x-u, y-v) G(u, v) du dv$$

$$= \iint f(x-u, y-v) \frac{1}{2\pi\sigma^2} e^{-\frac{(u^2+v^2)}{2\sigma^2}} du dv$$

$$= \iint f(x-u, y-v) \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{u^2}{2\sigma^2}} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{v^2}{2\sigma^2}} du dv$$

$$= \int \left[\int f(x-u, y-v) \underbrace{\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{u^2}{2\sigma^2}}}_{G_1(u)} du \right] \underbrace{\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{v^2}{2\sigma^2}}}_{G_2(v)} dv$$

$$= [f(x, y) * G_1(x)] * G_2(y)$$

$$= f(x, y) * [G_1(x) * G_2(y)]$$

$$\text{where } G_1(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}} \text{ and } G_2(y) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{y^2}{2\sigma^2}}$$

$$\text{Therefore } G(x, y) = G_1(x) * G_2(y),$$

which means that $G(x, y)$ must be separable.

Problem 3 (continued)

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b) Given: $G(x, y) \triangleq \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}}$

$$\nabla^2 G \triangleq \frac{\partial^2 G}{\partial x^2} + \frac{\partial^2 G}{\partial y^2}$$

$$\frac{\partial G}{\partial x} = \frac{-2x}{2\sigma^2} \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}} = \frac{-x}{2\pi\sigma^4} e^{-\frac{x^2+y^2}{2\sigma^2}}$$

$$\begin{aligned} \frac{\partial^2 G}{\partial x^2} &= \left(\frac{-1}{2\pi\sigma^4} e^{-\frac{x^2+y^2}{2\sigma^2}} \right) + \left(\left[\frac{-x}{2\pi\sigma^4} \right] \left[\frac{-2x}{2\sigma^2} \cdot e^{-\frac{x^2+y^2}{2\sigma^2}} \right] \right) \\ &= \left(\frac{x^2}{\sigma^4} - \frac{1}{\sigma^2} \right) \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}} \end{aligned}$$

Similarly,

$$\frac{\partial^2 G}{\partial y^2} = \left(\frac{y^2}{\sigma^4} - \frac{1}{\sigma^2} \right) \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}}$$

Combining the two parts:

$$\begin{aligned} \nabla^2 G &= \frac{\partial^2 G}{\partial x^2} + \frac{\partial^2 G}{\partial y^2} \\ &= \left(\frac{x^2+y^2}{\sigma^4} - \frac{2}{\sigma^2} \right) \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}} \\ &= \left(\frac{x^2+y^2}{\sigma^4} - \frac{2}{\sigma^2} \right) G(x, y) \end{aligned}$$

(As discussed in Section 7.2, this function can be split into two separable parts.)