## A simple implementation of the bilateral filter

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We'll do the implementation of the bilateral filter in its simplest form. The bilateral filter produces its output from an weighted combination of neighboring pixels. The spatial and intensities kernels will be Gaussian functions with standard-deviation  $\sigma_s$  and  $\sigma_i$  respectively. The window defining the neighborhood will be a square window.

• The image is defined on the finite grid

$$D = \{1, \dots, N\} \times \{1, \dots, M\} \subset \mathbb{N}^2. \tag{1}$$

- $p = (p_1, p_2) \in D$  is the pixel to be processed
- $u: D \to [0,1]$  is the image, so u(p) is the value of the pixel p
- $f_s(z) := \exp(\frac{-z^2}{2\sigma_s^2})$ ,  $z \in \mathbb{R}$ , is the Gaussian spatial kernel (it measures how far two pixels are),  $\sigma_s$  is a parameter
- $f_i(z) := \exp(\frac{-z^2}{2\sigma_i^2})$ ,  $z \in \mathbb{R}$ , is the Gaussian intensity kernel (it measures how two gray-values differ),  $\sigma_i$  is a parameter

We want to denoise the pixel  $p \in D$ , far enough from the boundaries of D. To this aim we define:

• the square of size  $(2w+1) \times (2w+1)$  (neighborhood) centered around p defined by

$$\Omega(p) := p + \{-w, \dots, w\}^2 = \left\{ y = (y_1, y_2) : y_1 \in \{p_1 - w, \dots, p_1 + w\}, \ y_2 \in \{p_2 - w, p_2 + w\} \right\}, \ (2)$$

for p far enough from the boundaries of D. The number  $w \in \mathbb{N}^+$  is a parameter.

The output of the bilateral filter is (we shall define the effective domain of p later on):

$$u_{denoised}(p) := \frac{1}{C} \sum_{y \in \Omega(p)} u(y) f_s(\|y - p\|_{\ell^2}) f_i(|u(y) - u(p)|)$$
(3)

$$=\frac{1}{C}\sum_{y_1=p_1-w}^{p_1+w}\sum_{y_2=p_2-w}^{p_2+w}u(y_1,y_2)\exp\left(-\frac{(y_1-p_1)^2+(y_2-p_2)^2}{2\sigma_s^2}\right)\exp\left(-\frac{[u(y_1,y_2)-u(p_1,p_2)]^2}{2\sigma_i^2}\right),$$

where the normalization constant is given by

$$C: = \sum_{y \in \Omega(p)} f_s (\|y - p\|_{\ell^2}) f_i (|u(y) - u(p)|)$$

$$= \sum_{y_1 = p_1 - w}^{p_1 + w} \sum_{y_2 = p_2 - w}^{p_2 + w} \exp\left(-\frac{(y_1 - p_1)^2 + (y_2 - p_2)^2}{2\sigma_s^2}\right) \exp\left(-\frac{[u(y_1, y_2) - u(p_1, p_2)]^2}{2\sigma_i^2}\right).$$

$$(4)$$

To denoise the entire image, loop over  $p \in D$  (and therefore  $\Omega(p)$ ) so that  $\Omega(p) \subset D$  (as usual, be careful with the boundaries) and apply the above formulas.

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## Part 1. Implementation

**Q1.** Compute the range of y - p when  $y \in \Omega(p)$ :

**Q2.** We recall that s and w are fixed. We consider the function  $\Omega(p) \ni y \mapsto f_s(\|y - p\|_{\ell^2})$ . Compute the range of  $f_s$ . Hint: Use **Q1.**.

**Q3.** Which part of u do we need to compute  $u_{denoised}(p)$ ?

**Q4.** From (3) compute the range of  $p_1$  and  $p_2$ , i.e., the  $p_1$  and  $p_2$  for which (3) is valid

**Q5.** We recall that in Matlab, the first index of a vector is 1. Rewrite (3) so that the summation indexes starts at 1.

**Q6.** Rewrite the formula you obtained for **Q5** using the function S et  $\tilde{u}$  given by

$$S: \{1, \dots, 2w+1\} \times \{1, \dots, 2w+1\} \ni (x_1, x_2) \mapsto \exp\left(-\frac{(x_1 - w - 1)^2 + (x_2 - w - 1)^2}{2\sigma_s^2}\right), \quad (5)$$

$$\tilde{u}: \{1, \dots, 2w+1\} \times \{1, \dots, 2w+1\} \ni (x_1, x_2) \mapsto u(x_1 + p_1 - w - 1, x_2 + p_2 - w - 1).$$
 (6)

**Q7.** Deduce the similar formula for C defined in (4).

**Q8.** List the objects needed to implement the program (Matrices of fixed size/varying size, matrices constant w.r.t the loop over p, real numbers, etc.)

**Q9.** Deduce the pseudo code of the program.

Part	2. Short analysis of the behavior of the bilateral filter
Q10.	What is the behavior of the bilateral filter when $\sigma_i \to +\infty$ ?

**Q11.** What happens when  $\sigma_i \to 0$ ?

**Q12.** What happens when  $\sigma_s \to 0$ ?

**Q13.** What happens when  $\sigma_s \to +\infty$ ?

