

Assignment 2: CS 663, Fall 2024

Due: 6th September before 11:55 pm

Remember the honor code while submitting this (and every other) assignment. You may discuss broad ideas with other students or ask me for any difficulties, but the code you implement and the answers you write must be your own. We will adopt a zero-tolerance policy against any violation.

Submission instructions: Follow the instructions for the submission format and the naming convention of your files from the submission guidelines file in the homework folder. Please see `assignment2.zip` in the homework folder. For all the questions, write your answers and scan them, or type them out in word/Latex. In either case, create a separate PDF file. The last two questions will also have code in addition to the PDF file. Once you have finished the solutions to all questions, prepare a single zip file and upload the file on moodle before 11:55 pm on 6th September. **Only one student per group should submit the assignment.** We will not penalize submission of the files till 10 am on 7th September. **No assignments will be accepted after this time.** Please preserve a copy of all your work until the end of the semester. **Your zip file should have the following naming convention:** RollNumber1.RollNumber2.RollNumber3.zip for three-member groups, RollNumber1.RollNumber2.zip for two-member groups and RollNumber1.zip for single-member groups.

1. Consider a clean image $I(x, y)$ which gets corrupted by additive noise randomly and independently from a zero mean Gaussian distribution with standard deviation σ . Derive an expression for the PDF of the resulting noisy image. Assume continuous-valued intensities. What fundamental operation from image processing does the expression for the PDF resemble? How will you modify the expression for the resultant PDF if the noise were uniformly distributed from $-r$ to $+r$ where $r > 0$? [5+3+2=10 points]
2. Consider a 1D convolution mask given as (w_0, w_1, \dots, w_6) . Express the convolution of the mask with a 1D image f as the multiplication of a suitable matrix with the image vector f . What are the properties of this matrix? What could be a potential application of such a matrix-based construction? [10 points]
3. Prove or disprove: (a) The Laplacian mask with a -8 in the center (see class slides) is a separable filter. (b) The Laplacian mask with a -4 in the center (see class slides) can be implemented entirely using 1D convolutions. [5+5=10 points]
4. Consider an image $I(x, y)$. Argue using a 1D example that adding $\alpha \nabla^2 I(x, y)$ to $I(x, y)$ where $\alpha > 0, \alpha \approx 0$ is a small constant, causes blurring of the edges. What would happen if the operation $I(x, y) \leftarrow I(x, y) + \alpha \nabla^2 I(x, y)$ were run for a very large number of iterations? Why? On the other hand, what would happen if the operation $I(x, y) \leftarrow I(x, y) - \alpha \nabla^2 I(x, y)$ were run for a very large number of iterations? Why? [7+4+4=15 points]
5. Consider a 1D ramp image of the form $I(x) = cx + d$ where c, d are scalar coefficients. Derive an expression for the image J which results when I is filtered by a zero-mean Gaussian with standard deviation σ . Derive an expression for the image that results when I is treated with a bilateral filter of parameters σ_s, σ_r . (Hint: in both cases, you get back the same image.) Ignore any border issues, i.e. assume the image had infinite extent. [10 points]
6. Prove that the Laplacian operator is rotationally invariant. For this consider a rotation of the coordinate system from (x, y) to $u = x \cos \theta - y \sin \theta, v = x \sin \theta + y \cos \theta$, and show that $I_{xx} + I_{yy} = I_{uu} + I_{vv}$ for any image I . Prove that the second directional derivative of an image $I(x, y)$ in the direction of its

gradient vector (*i.e.* in the direction $(\frac{I_x}{\sqrt{I_x^2 + I_y^2}}, \frac{I_y}{\sqrt{I_x^2 + I_y^2}})$) is given by $\frac{I_x^2 I_{xx} + 2I_x I_y I_{xy} + I_y^2 I_{yy}}{I_x^2 + I_y^2}$. Note

that $I_x = \frac{\partial I}{\partial x}$, $I_{xx} = \frac{\partial^2 I}{\partial x^2}$. Using this information, write down the expression for the second directional derivative of $I(x, y)$ in the direction **perpendicular** to its gradient vector and justify your answer. Note that the first directional derivative of $I(x, y)$ in a direction v is given by $\nabla I(x, y) \cdot v$. [6+6+3 = 15 points]

7. Consider the two images in the homework folder ‘barbara256.png’ and ‘kodak24.png’. Add zero-mean Gaussian noise with standard deviation $\sigma = 5$ to both of them. Implement a bilateral filter and show the outputs of the bilateral filter on both images for the following parameter configurations: $(\sigma_s = 2, \sigma_r = 2)$; $(\sigma_s = 0.1, \sigma_r = 0.1)$; $(\sigma_s = 3, \sigma_r = 15)$. Comment on your results in your report. Repeat when the image is corrupted with zero-mean Gaussian noise of $\sigma = 10$ (with the same bilateral filter parameters). Comment on your results in your report. For the bilateral filter implementation, write a MATLAB function `mybilateralfilter.m` which takes as input an image and parameters σ_r, σ_s . Implement your filter using at the most two nested for-loops for traversing the image indices. For creating the filter, use functions like `meshgrid` and vectorization for more efficient implementation. Include all image outputs as well as noisy images in the report. [15 points]
8. Implement local histogram equalization of sizes $7 \times 7, 31 \times 31, 51 \times 51, 71 \times 71$ on the images ‘LC1.jpg’ and ‘LC2.jpg’ from the homework folder. Comment on your results in your report and compare it to global histogram equalization, which you can use from the image processing toolbox of MATLAB. Point out regions where the local method produces better local contrast than the global histogram equalization. [15 points]