Homework 2

Introduction to Image Analysis

Handout: 6th, April 2022 Handin: May 11th, 2022, 13:15

Instructions

The following homework is an opportunity for you to have hands on experience manipulating and working with images and the material covered in class.

Your hand-in will consist of a .zip archive named hw2_firstName_lastName.zip containing the following:

- Python source files named hw2_exY_firstName_lastName.py, where Y is the exercise number
- All necessary files to run the above.
- Your report named hw2_firstName_lastName.pdf.

Your archive must be uploaded on ILIAS before the deadline.

Code [70 Points]

Code templates are provided to help you get started in solving the exercises. In the typical case, you will fill-in the missing function bodies and code blocks. Note that you may also make your own code from scratch.

IMPORTANT: In general, if you are not able to produce a functional code (i.e. your script crashes), comment out the concerned part, and if necessary give a short analysis in your report. Scripts that do not run will be penalized.

Report [30 Points]

Along with the code, you are asked to provide a short report. Comment on all the questions on the report, show your results, including figures of all requested plots, and briefly explain what you did to obtain them. If you encountered problems and did not manage to finish the exercise, explain here what you did. On ILIAS you will find a LATEX template to get started. Note that the use of LATEX is NOT mandatory.

1 Linear Filtering - [25 Points]

1.1 [2 Points]

Write a function that returns a box filter of size $[n \times n]$. Make sure the filter values sum up to 1. If you are using the provided template you may fill in the boxfilter(n) function.

HINT: This can be done as a simple one-line function.

1.2 [5 Points]

Implement a 2D convolution function between an image $[m \times n]$ and a filter $[k \times l]$. DO NOT USE any built-in convolution functions (such as from scipy or numpy). You should code your own version of the convolution, valid for both 2D and 1D filters.

If you are using the provided template you may fill in the myconv2(image,filt) function.

IMPORTANT: The function should return a full convolution, meaning the output should be of size $(m + k - 1) \times (n + l - 1)$ Use any border filling approach you choose to ensure values at the border. For example, you could assume 0 for values outside the image (zero padding).

1.3 [2 Points]

In your script, create a boxfilter of size 11 and convolve this filter with your image. Show your result.

1.4 [4 Points]

Write a function that returns a 1D Gaussian filter for a given value of sigma. The filter should be a vector of length filter_length, if filter_length is odd, filter_length + 1 otherwise. Each value of the filter can be computed from the Gaussian function, $e^{-\frac{x^2}{2*\sigma^2}}$, where x is the distance of an array value from the center. This formula for the Gaussian ignores a constant factor, so you should normalize the values in the filter so they sum to 1.

If you are using the provided template you may fill in the gauss1d(sigma, filter_length) function.

HINTS: For efficiency and compactness, it is best to avoid "for" loops. One way to do this is to first generate a vector of values for x, for example [-3 -2 -1 0 1 2 3] for a filter with length 7. These can then be used to calculate the Gaussian value corresponding to each element.

1.5 [2 Points]

Write a function that returns a 2D Gaussian filter. Remember that a 2D Gaussian can be formed by convolution of a 1D Gaussian with its transpose. Use your function from 1.2 to perform the convolution.

Display a 2D gaussian filter with sigma=3 and size 11x11.

If you are using the provided template you may fill in the gauss2d(sigma, filter_size) function.

1.6 [3 Points]

Implement 2D gaussian filtering of an image.

Convolve the image with a gaussian filter with a sigma of 3 and display the result.

If you are using the provided template you may fill in the gconv(image, sigma) function.

1.7 [3 Points]

Convolution with a 2D Gaussian filter is not the most efficient way to perform Gaussian convolution with an image. In a few sentences, explain how this could be implemented more efficiently and why this would be faster.

Answer this question in your report.

HINT: How can we use 1D Gaussians?

1.8 [4 Points]

To show this fact create a plot of filter size vs. computation time. In this plot, show in one color (e.g. blue) what this size-time relation is when using 2D box filter of increasing size (3,100) with a step of 5. Repeat the same for your solution to (1.7) and show it in another color.

2 Finding edges - [20 Points]

2.1 [2 Points]

Begin by defining a derivative operator of your choosing. For example, dx = [-1, 0, 1]. Using your functions from Exercise 1, convolve dx with a Gaussian filter with sigma = 1. Repeat the same for dy.

2.2 [4 Points]

Using dx and dy, construct an edge magnitude image. That is, for every pixel in the image, assign the magnitude of gradients. Also calculate for each pixel the gradient orientation.

Display the magnitude image next to the original one.

If you are using the provided template you may fill in the create_edge_magn_image(image, dx, dy) function.

2.3 [6 Points]

You will now produce edge images of particular directions. An edge image is a binary image. In this case, you will produce 8 such images. For $i=0,\ldots,7$, an edge image e_i should have values 0 except if a pixel has an edge of magnitude greater than threshold r and has an orientation in the interval $(\frac{2i\pi}{8} - \frac{\pi}{8}, \frac{2i\pi}{8} + \frac{\pi}{8})$. In this case, the pixel should have value 255. Please find a value for r which is suitable for your image and describe your reasoning for this threshold in the report.

If you are using the provided template you may fill in the make_edge_map(image, dx, dy) function.



Figure 1: Edge maps on circle.

Show the result of these edge images by concatenating all of them with the original and magnitude edge image.

HINT: To verify your code, run it on image circle.jpg. You should generate edge images that cover the extent of the perimeter of the circle, as shown on figure 1.

2.4 [8 Points]

Implement edge non-max suppression on the magnitude edge image, as in canny edge detection, in order to reduce the thickness of the edge response you observe.

If you are using the provided template you may fill in the edge_non_max_suppression(img_edge_mag, edge_maps) function.

Display the result showing the original image, the magnitude edge image and the result of the non-maximum suppression.

3 Corner detection - [25 Points]

For this exercise you can either use your functions from exercise 1, or scipy.signal.convolve().

3.1 [15 Points]

Write a function which computes the harris corner for each pixel in the image. The function should return the R response at each location of the image.

If you are using the provided template you may fill in the myharris(image, w_size, sigma, k) function.

HINT: You may have to play with different parameters to have appropriate R maps. Try Gaussian smoothing with sigma = 0.2; Gradient summing over a 5x5 region around each pixel and k = 0.1.

3.2 [2 Points]

Evaluate your function from the previous question on the chessboard image. Show the result.

3.3 [2 Points]

Repeat the same as (3.2), but this time rotate the chessboard image by 45 degrees (in either direction). Show the result.

You can use the built in scipy.ndimage.rotate() function.

3.4 [2 Points]

Repeat the same as (3.2), but this time downscale the chessboard image by a factor of a half. Show the result.

You can use the built in scipy.misc.imresize() function.

3.5 [4 Points]

Looking at the results from (3.2), (3.3) and (3.4) what can we say about the properties of Harris corners? What is maintained? What is it invariant to? Why is that the case?

Answer this question in your report.