

CV - Assignment 2- Practical- Report

Q1) Image Smoothing

Applied a 3*3 and 5*5 Averaging filter:

1/9	1/9	1/9
1/9	1/9	1/9
1/9	1/9	1/9

1/25	1/25	1/25	1/25	1/25
1/25	1/25	1/25	1/25	1/25
1/25	1/25	1/25	1/25	1/25
1/25	1/25	1/25	1/25	1/25
1/25	1/25	1/25	1/25	1/25

Code and Algo Description:

There are 4 functions:

showImage(image, title): A simple matplotlib function to display the images.

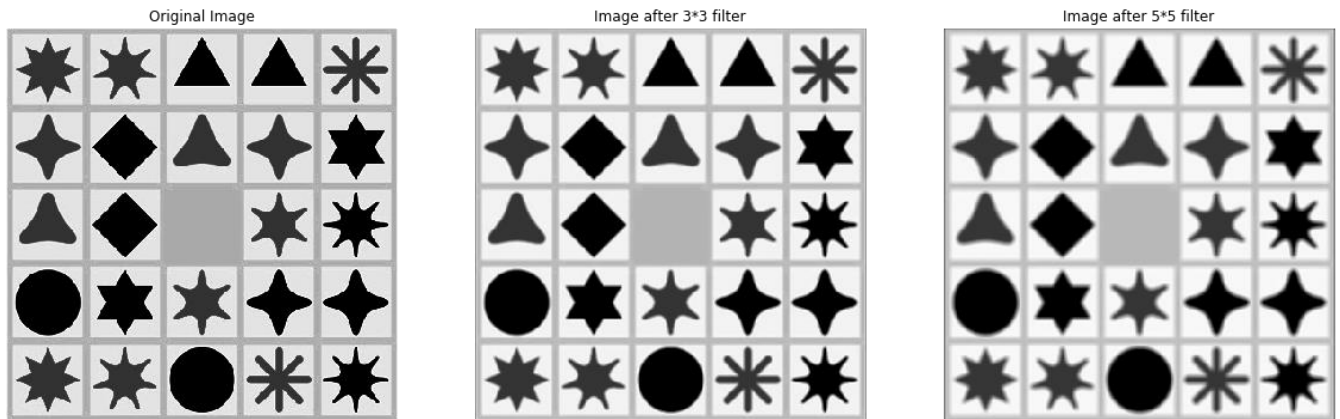
padding(k, original_image): Used for padding an image for k*k filter size. Padding has been done with 0s on all four sides of the image. (I do have another function for padding for a m*n filter as well)

sliding_window(image, stepSize, windowSize): This function simply slides across a given image using the window_size specified. It returns a generator which is then used by the next averaging function.

averaging(k, original_image, padded_image, filter_mask): This function takes in the generator passed by the sliding_window function and takes only selected matrices passed to it. It will then multiply the window with the filter_mask (numpy matrix multiplication) and then sum up all the values and use that as the new value for the final image.

Name: Udita Gupta
Net ID: ung200
N#: N17237066

Output:



Q2) Edge Detection

Functions used:

showImage(image, title)

padding(k, original_image)

sliding_window(image, stepSize, windowSize)

averaging(k, original_image, padded_image, filter_mask)

Steps:

For this we first have to find dx and dy of the original image.
I have used Sobel Operators for getting dx dy.

Dx:

Filter used:

-1	0	1
-2	0	2
-1	0	1

Dy:

Filter used:

-1	-2	-1
0	0	0
1	2	1

For Edge Map:

Compute the following: (We have dx and dy values already)

$$\|\nabla f\| = \sqrt{\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2}$$

For Orientation Map:

Compute the following: (We have dx and dy values)

$$\Theta = \tan^{-1}(-dx/dy)$$

Output:

Original Image



Image after 3*3 filter



Image dx

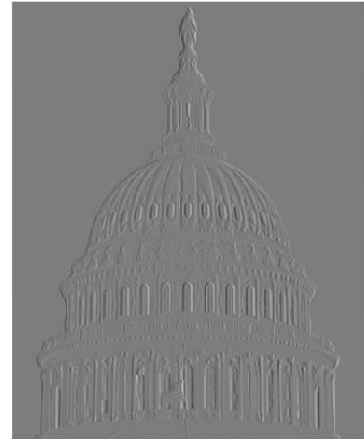


Image dy

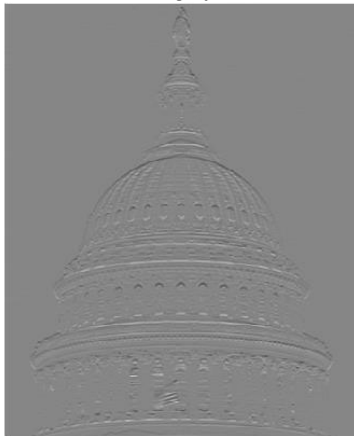
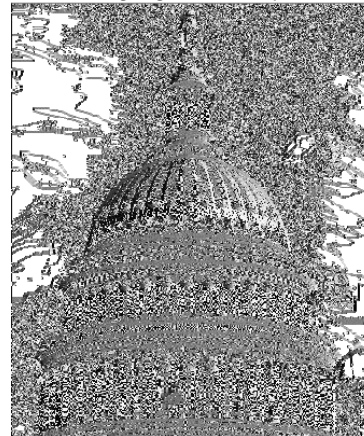


Image Edge Map



Image Edge Orientation Map



Q3) Template Matching

Functions:

zero_mean(original_matrix): This function is simply used for calculating the zero-mean matrix. That is, subtract the mean value of the matrix from each value in the matrix.

padding_not_square(m, n, original_image): This is same as before padding function but this time it's for a $m \times n$ filter. Padding has been done with 0s on all four sides of the image.

sliding_window(image, stepSize, windowSize): This is same as before.

averaging_not_square(m, n, original_image, padded_image, w_matrix): - This is the same as previous averaging function with a few new additions.

- Like mentioned in the template matching Algo in the question: the given sliding window is first re-computed using the zero-mean function.
- Then the two matrices (zero-mean window and template) are convoluted as before forming the **g** correlation matrix.
- It also computes the cross-correlation matrix **c** as mentioned in the algorithm by dividing each value of the **g** matrix by the magnitudes of the template matrix and window matrix.
- This function returns back both the **g** and **c** matrices.

find_threshold(matrix): This function is used for finding the threshold. We first take the **c** matrix returned from the above function and flatten it.

- Sort the list in ascending order
- Take the last 100 values (the max intensities)
- take an average of these 100 values as the threshold.

apply_threshold(threshold, matrix): This function takes in the threshold value and the matrix it has to apply thresholding on.

- Matrix values above threshold are kept and matrix values below the threshold are made as 0.
- We get a black image with white spots where the peaks were. Basically where we got maximum correlation between the template and the image.

Algorithm used:

- Take the input image **i**, and template, **t**
- Calculate zero-mean matrix of the template, let's call it **zero_mean_t**.
- Pad the image **i**

- Get all the windows (matrices) from the image i
- Calculate zero-mean of each window, lets call it w
- Convolve w and $zero_mean_t$, lets call it, g
- Multiply each value in matrix g by $(1/(||w|| \cdot zero_mean_t \cdot ||))$, lets call it, c
- Get both g and c matrices and plot them
- Apply thresholding to the c matrix and get the maximum peaks.
- You can also try, direct Laplacian on matrix c .
- One more method is to smooth the matrix c and then apply laplacian to it, followed by thresholding. This basically means, LoG on matrix c and then thresholding.
- Plot all the results.

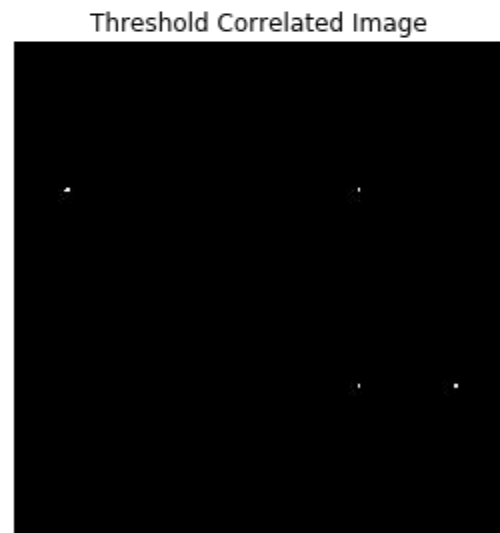
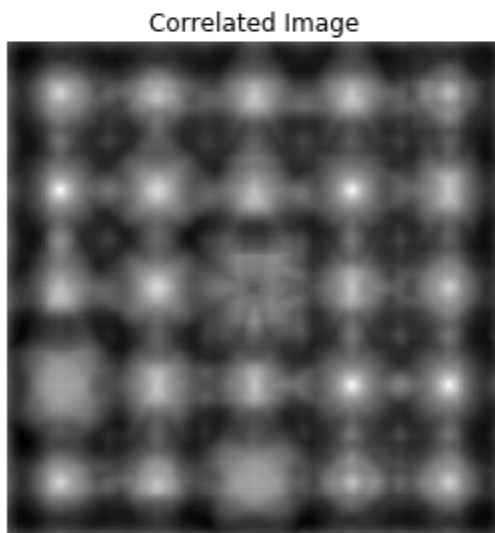
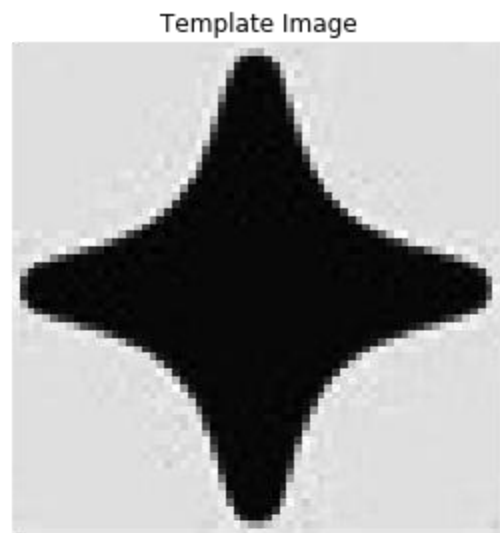
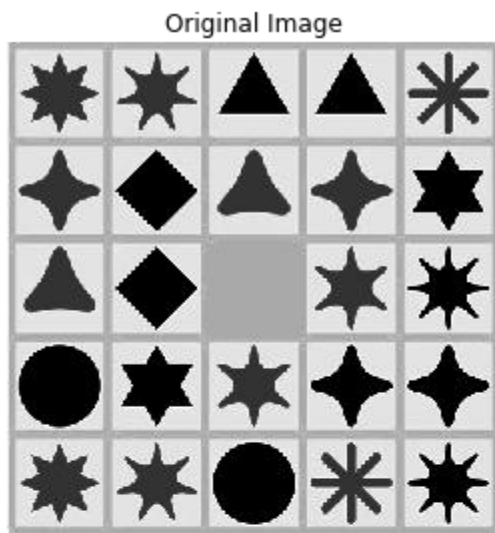
Laplacian:

0	-1	0
-1	4	-1
0	-1	0

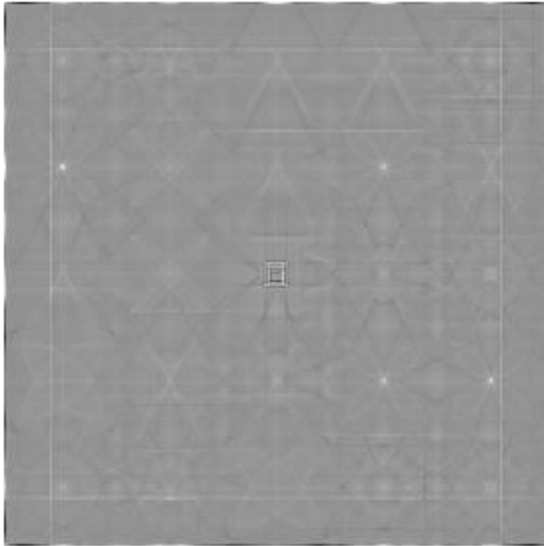
LoG: (Have used positive values to get bright points in the image instead of dark points)

0	0	1	0	0
0	1	2	1	0
1	2	16	2	1
0	1	2	1	0
0	0	1	0	0

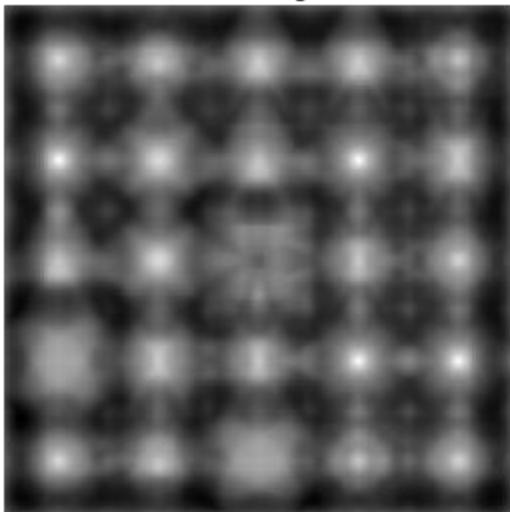
Output :



Laplacian Image



LoG Image



LoG Thresholded Image

