

# QUESTION 8

## CS663 (DIGITAL IMAGE PROCESSING) ASSIGNMENT 2

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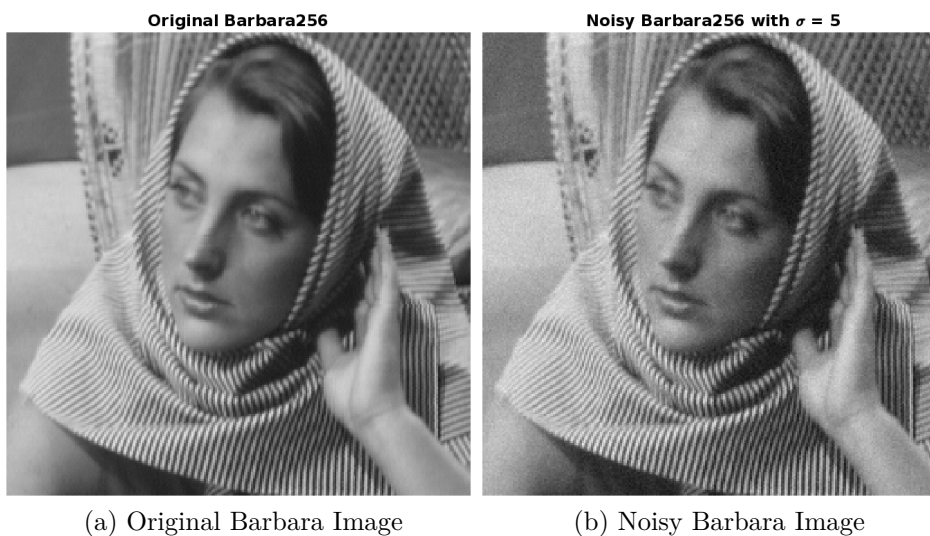
## Question 8

### Problem 1

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  $\sigma_r, \sigma_s$ . 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]

### SECTION 1

## Barbara256 with Gaussian Noise ( $\sigma = 5$ )



**Figure 1.** (b) shows Noisy Barbara with Gaussian Noise ( $\mu = 0, \sigma = 5$ )

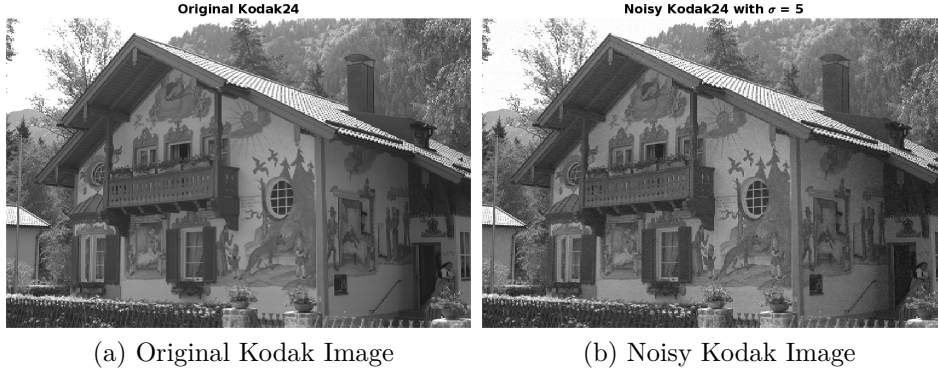


**Figure 2.** Barbara Images after applying Bilateral Filter

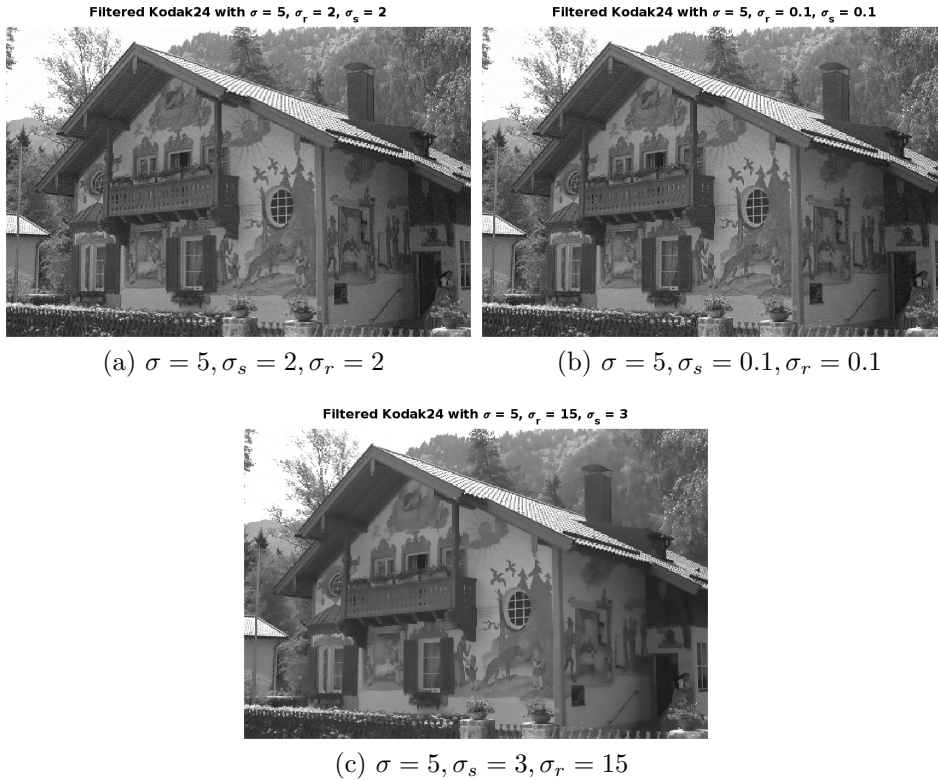
From the above results, we observe that the image gets more blurred when  $\sigma_s$  and  $\sigma_r$  are increased for the bilateral filter. When  $\sigma_s$  increases, a larger and larger neighborhood of values around a pixel  $(x, y)$  will contribute to the averaging (more noise reduction but possible contribution from dissimilar regions).

Also, when  $\sigma_r$  is more, then the blurring is even more because it controls the amount of different intensities taken around a pixel for averaging. Features or edges with intensity difference less than  $\sigma_r$  become blurred, and others remain preserved. At higher values, the bilateral filter begins to resemble a Gaussian filter.

## SECTION 2

Kodak24 with Gaussian Noise ( $\sigma = 5$ )

**Figure 3.** (b) shows Noisy Kodak with Gaussian Noise ( $\mu = 0, \sigma = 5$ )



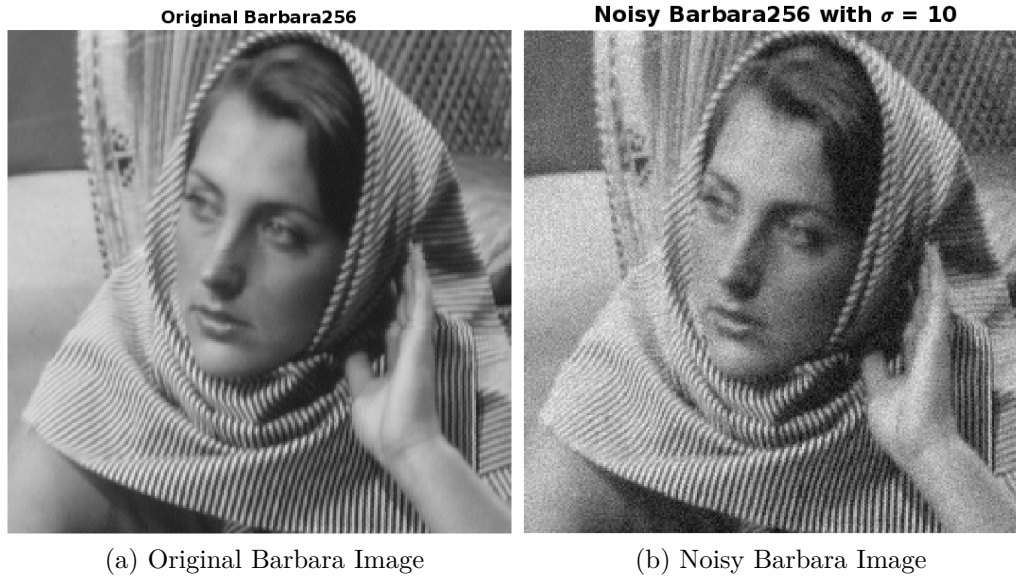
**Figure 4.** kodak Images after applying Bilateral Filter

Similar results as explained before in barbara image are obtained for kodak image also in terms of varying  $\sigma_r$  and  $\sigma_s$

## SECTION 3

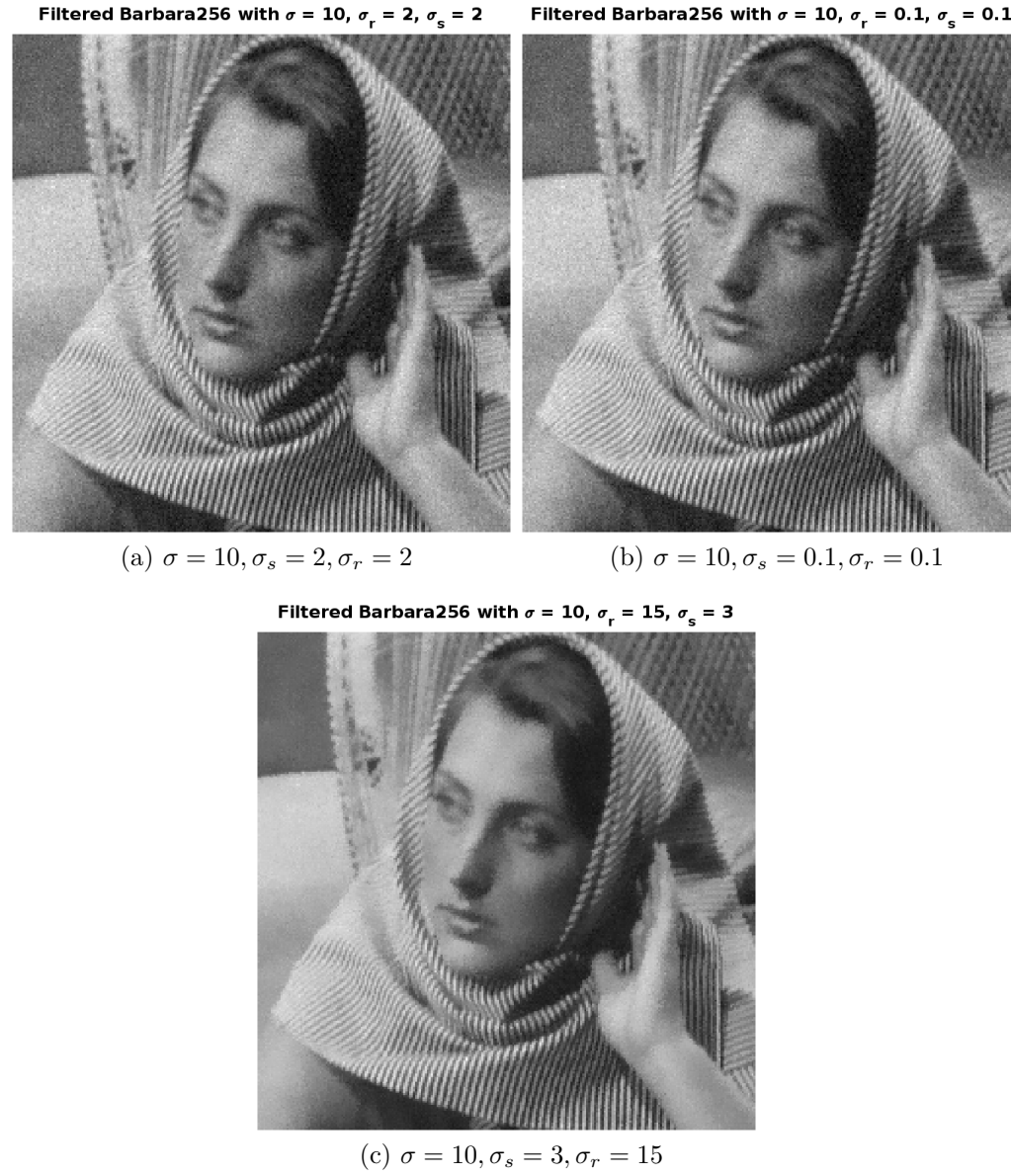
**Barbara256 with Gaussian Noise ( $\sigma = 10$ )**

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**Figure 5.** (b) shows Noisy Barbara with Gaussian Noise ( $\mu = 0, \sigma = 10$ )

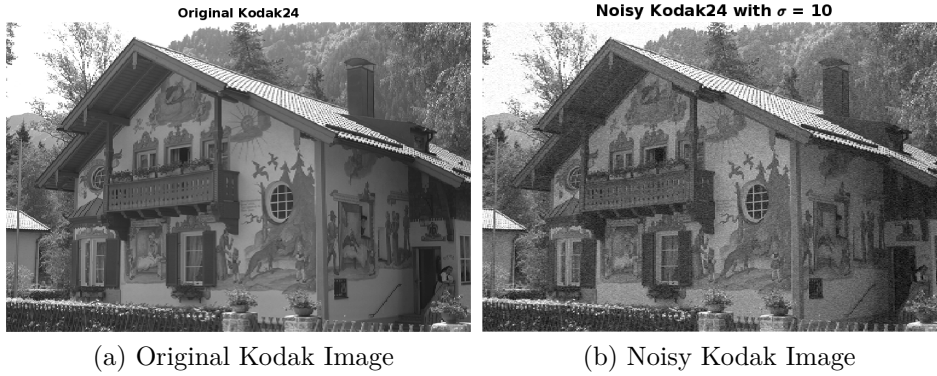
Now, the value of standard deviation of the Gaussian Noise added is increased from 5 to 10. We repeated the same processes and observed that as  $\sigma$  increased, the extent of noise added to the image increased because the Gaussian curve widens on increasing  $\sigma$ , and hence more error is added to the image intensities. The trend in filtering results were similar, that is  $\sigma_r = 3$ ,  $\sigma_s = 15$  case provided better filtering than other ones, but with some blur.



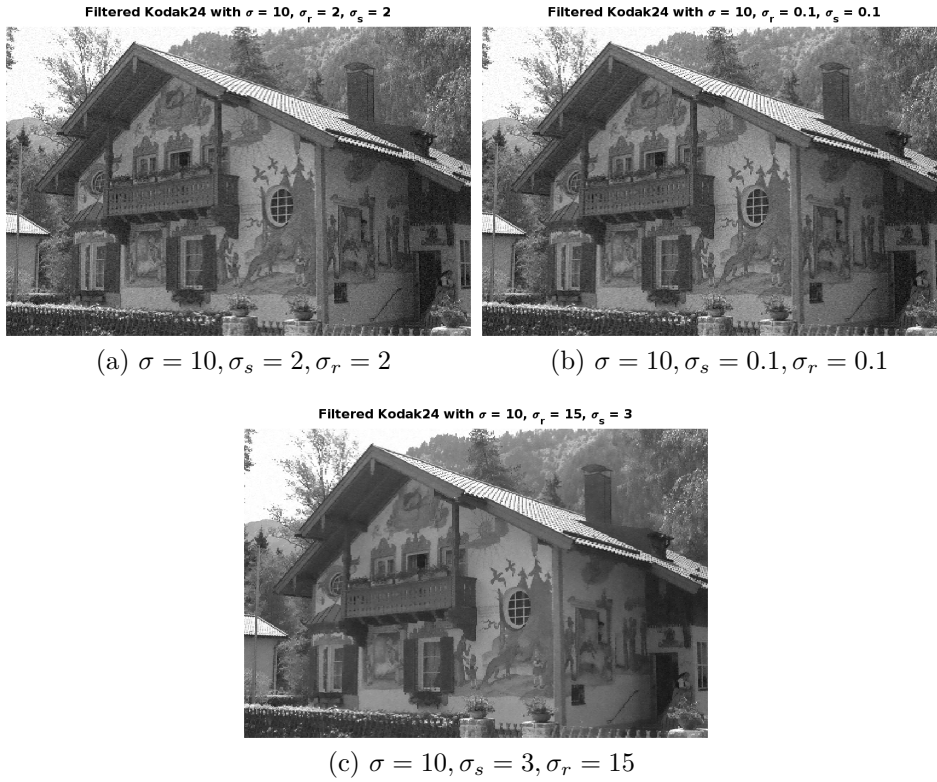
**Figure 6.** Barbara Images after applying Bilateral Filter



## SECTION 4

Kodak24 with Gaussian Noise ( $\sigma = 10$ )

**Figure 7.** (b) shows Noisy Kodak with Gaussian Noise ( $\mu = 0, \sigma = 10$ )



**Figure 8.** kodak Images after applying Bilateral Filter