



Computer Vision Assignment 2
Kai-Po Lin (kl72); Department of Computer Science

[Part 1]

Exercise 1:

1. Original Marvel Superhero Image (before cropping).



IronMan.jpg (https://www.google.com/url?sa=i&url=https%3A%2F%2Fsquare-enix-games.com%2Fen_GB%2Fnews%2Fmarvels-avengers-pro-tips-iron-man&psig=AOvVaw0CgnMcB1_Gsd7E83Kg2pKW&ust=1643226482615000&source=images&cd=vfe&ved=0CAsQjRxqFwoTCOi-xpLWzfUCFQAAAAAdAAAAABAD)

2. Cropped Head.

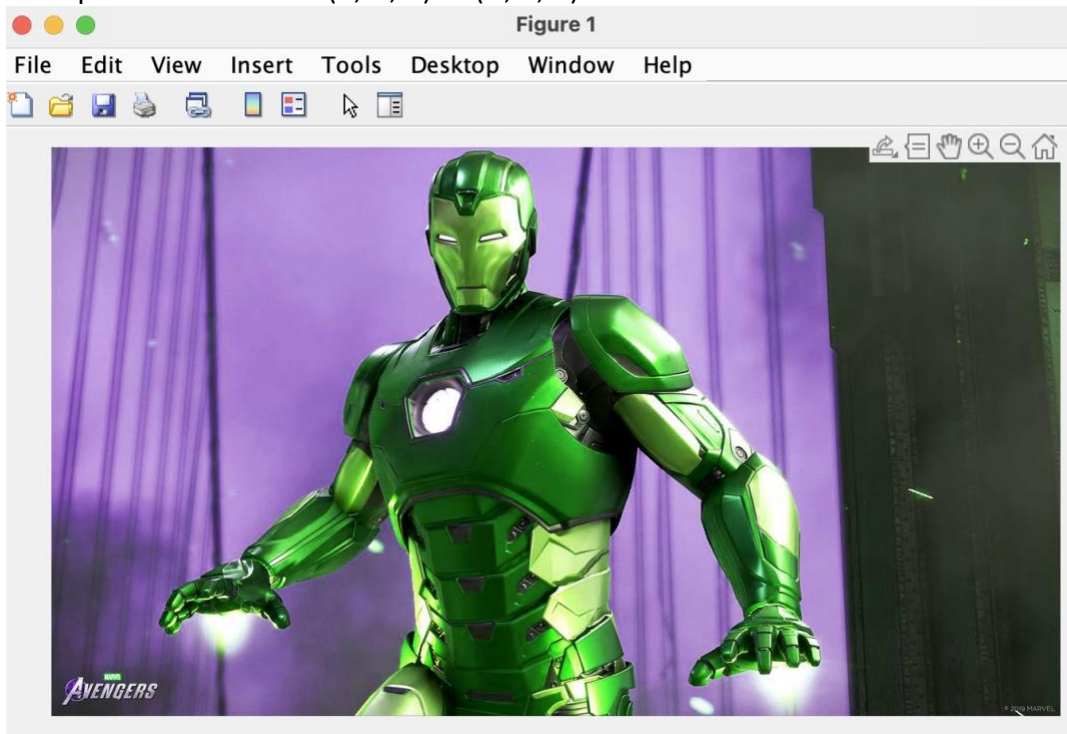


head.png

3. Green Component of Head.

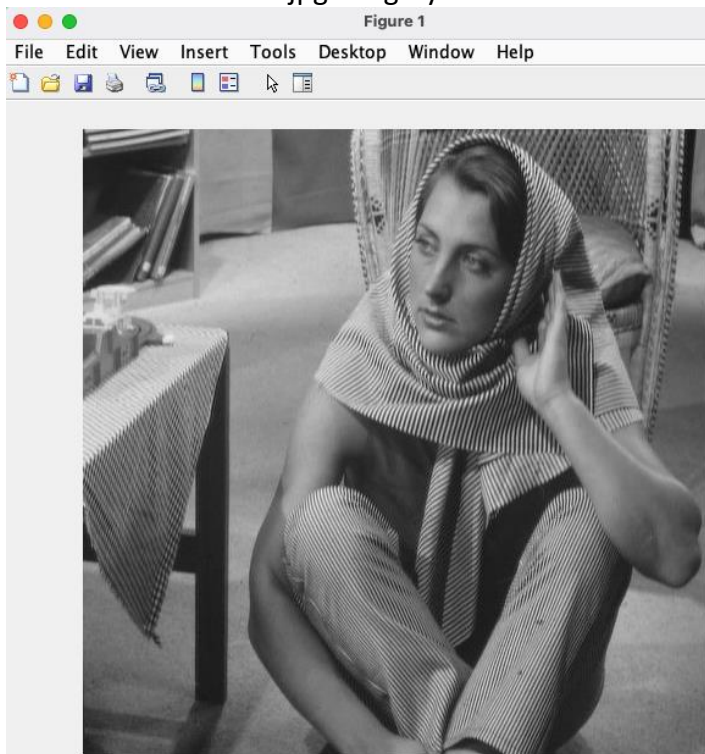


4. Transpose channel from (R, G, B) to (G, R, B).

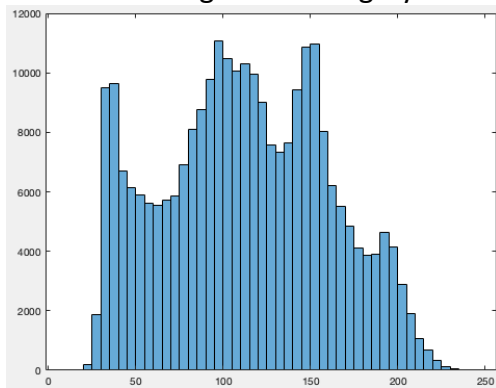


Exercise 2:

1. Convert "barbara.jpg" to gray scale.



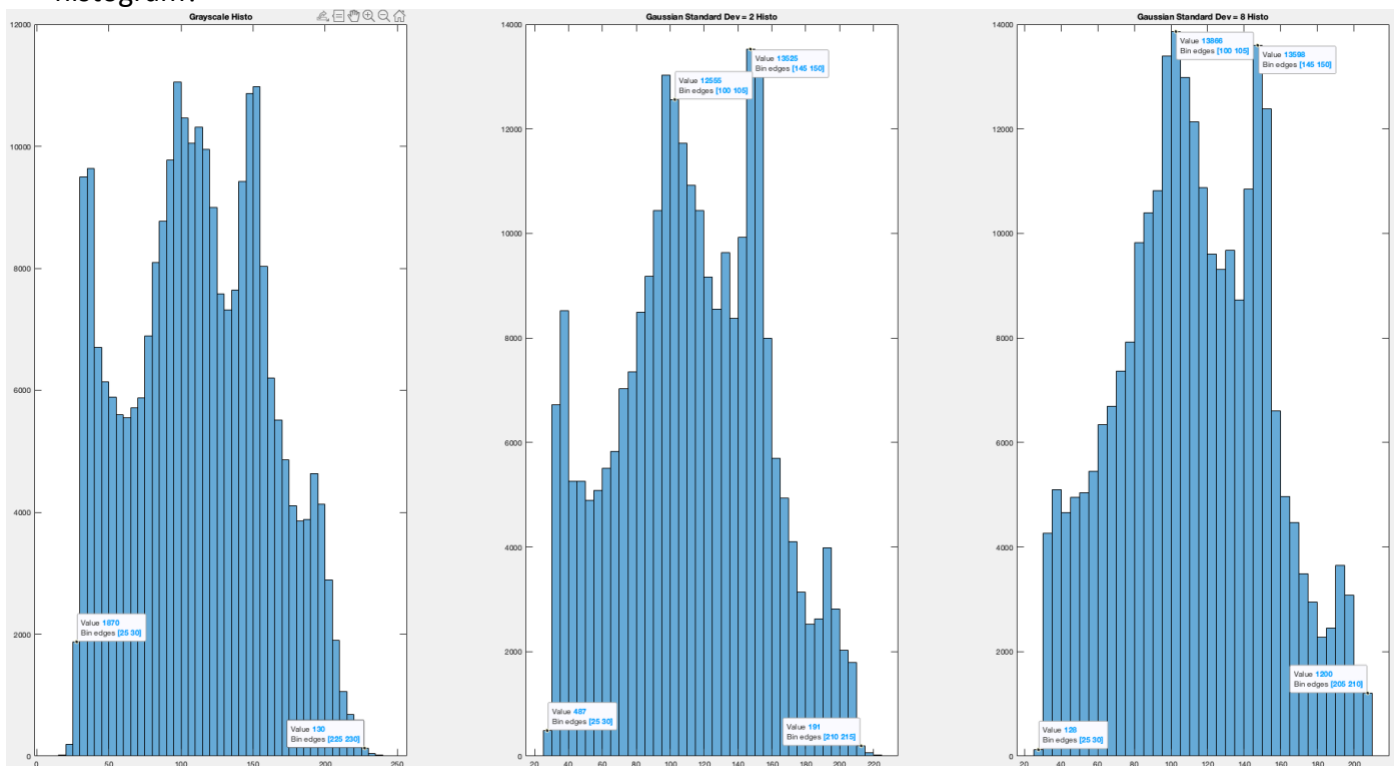
2. Plot a histogram of the gray scale image with bin-size of 5 in intensity.



3. Blur the gray scale image by using Gaussian filters of size 15×15 with standard deviations 2 and 8.



4. Plot histograms for the blurred images. How do they differ from each other and the original histogram?

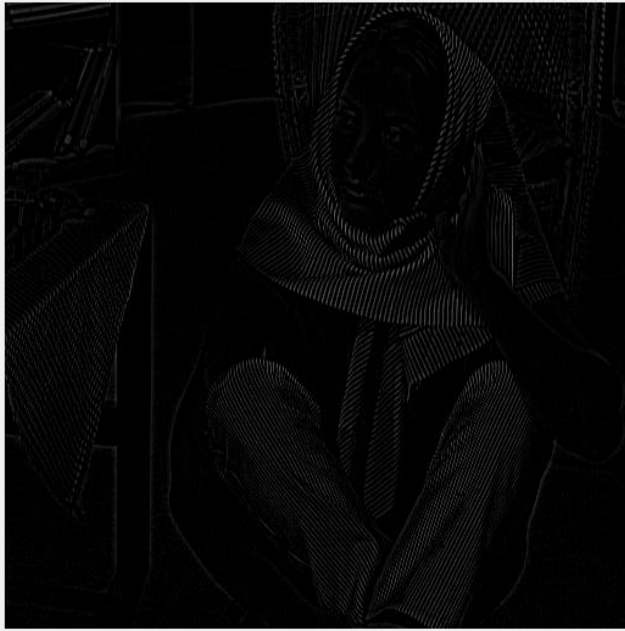


According to the histograms above, we can find out that as the standard deviation of the Gaussian filter increases (the original grayscale image can be treated as a 0-standard deviation), **the intensities of the image are more likely to centralize**. That means that a Gaussian filter with a higher standard deviation is more likely to average the intensity of data, causing the image to blur more intensively.

Intensity (Approximately)

- ⇒ Original Image: Start from intensity=20 with value 198 to intensity=235 with value 45
- ⇒ 2 Standard Deviation: Start from intensity=25 with value 487 to intensity=215 with value 191
- ⇒ 8 Standard Deviation: Start from intensity=25 with value 128 to intensity=210 with value 1200

5. Subtract the blurred image obtained using the filter with standard deviation of 2 from the original gray scale image.



6. Threshold the resultant image at 5% of its maximum pixel value and display the final image.



[Part 2]

Filtering:

Filter the following matrices I_1 and I_2 without using built-in MATLAB functions.

$$I_1 = \begin{bmatrix} 120 & 110 & 90 & 115 & 40 \\ 145 & 135 & 135 & 65 & 35 \\ 125 & 115 & 55 & 35 & 25 \\ 80 & 45 & 45 & 20 & 15 \\ 40 & 35 & 25 & 10 & 10 \end{bmatrix} \quad I_2 = \begin{bmatrix} 125 & 130 & 135 & 110 & 125 \\ 145 & 135 & 135 & 155 & 125 \\ 65 & 60 & 55 & 45 & 40 \\ 40 & 35 & 40 & 25 & 15 \\ 15 & 15 & 20 & 15 & 10 \end{bmatrix}$$

(I1, I2 used zero padding before filtering)

1. $1/3 [1 \ 1 \ 1]$

(1) Filter1 on I1

	1	2	3	4	5
1	76.6667	106.6667	105	81.6667	51.6667
2	93.3333	138.3333	111.6667	78.3333	33.3333
3	80	98.3333	68.3333	38.3333	20
4	41.6667	56.6667	36.6667	26.6667	11.6667
5	25	33.3333	23.3333	15.0000	6.6667

(2) Filter1 on I2

	1	2	3	4	5
1	85	130	125.0000	123.3333	78.3333
2	93.3333	138.3333	141.6667	138.3333	93.3333
3	41.6667	60.0000	53.3333	46.6667	28.3333
4	25	38.3333	33.3333	26.6667	13.3333
5	10	16.6667	16.6667	15.0000	8.3333

2. $1/3 [1; 1; 1]$

(1) Filter2 on I1

	1	2	3	4	5
1	88.3333	81.6667	75	60.0000	25
2	130	120	93.3333	71.6667	33.3333
3	116.6667	98.3333	78.3333	40	25
4	81.6667	65	41.6667	21.6667	16.6667
5	40	26.6667	23.3333	10	8.3333

(2) Filter2 on I2

	1	2	3	4	5
1	90	88.3333	90	88.3333	83.3333
2	111.6667	108.3333	108.3333	103.3333	96.6667
3	83.3333	76.6667	76.6667	75	60
4	40	36.6667	38.3333	28.3333	21.6667
5	18.3333	16.6667	20	13.3333	8.3333

3. $\frac{1}{9} [1 \ 1 \ 1; 1 \ 1 \ 1; 1 \ 1 \ 1]$

(1) Filter3 on I1

matrix3_1					
5x5 double					
	1	2	3	4	5
1	56.6667	81.6667	72.2222	53.3333	28.3333
2	83.3333	114.4444	95	66.1111	35
3	71.6667	97.7778	72.2222	47.7778	21.6667
4	48.8889	62.7778	42.7778	26.6667	12.7778
5	22.2222	30	20	13.8889	6.1111

(2) Filter3 on I2

matrix2_2					
5x5 double					
	1	2	3	4	5
1	59.4444	89.4444	88.8889	87.2222	57.2222
2	73.3333	109.4444	106.6667	102.7778	66.6667
3	53.3333	78.8889	76.1111	70.5556	45
4	25.5556	38.3333	34.4444	29.4444	16.6667
5	11.6667	18.3333	16.6667	13.8889	7.2222

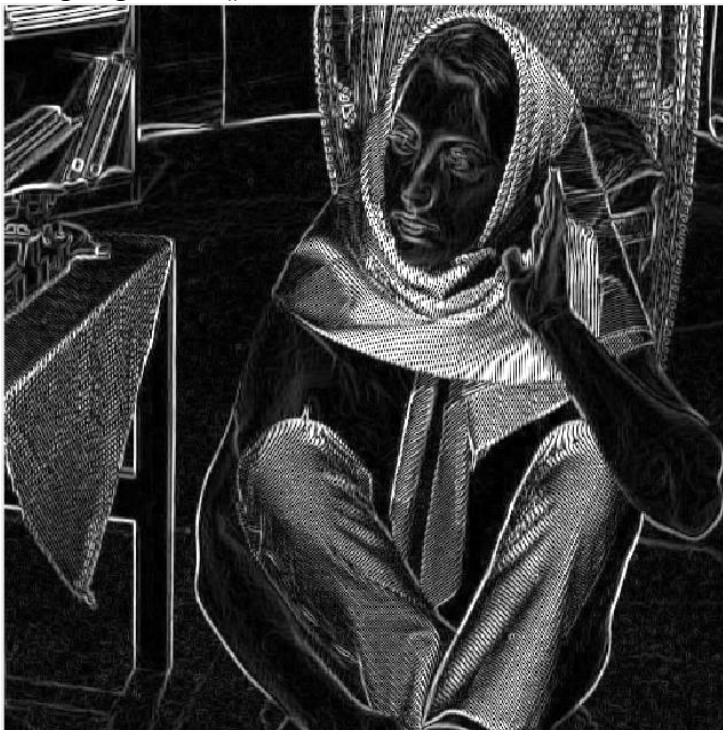
Apply following filters on the gray scale image of barbara from Part I.

1. Central difference Gradient filter
2. Sobel filter
3. Mean filter
4. Median filter

Note: You can use fspecial, imfilter functions in Matlab to accomplish the above and below tasks. Do not use imgradient.

1. Central difference Gradient filter

Using imgradient() => Should be the correct answer



Using $x_filter = [-1, 0, 1]/2$, $y_filter = [-1; 0; 1]/2$, $gradient_image = \sqrt{x_image.^2 + y_image.^2}$;
=> **Mathematically correct, but the output image seems too dark.**



Using $x_filter = [-1, 0, 1]/2$, $y_filter = [-1; 0; 1]/2$, $gradient_image = \sqrt{x_image.^2 + y_image.^2}$;
=> **Mathematically wrong, but the output image is more similar to the expected image that result from `imgradient()`**



2. Sobel filter



3. Mean filter



4. Median filter



Smoothing:

Original Image:



1. Averaging filters of sizes 2x2,4x4,8x8,16x16.

Average 2x2



Average 4x4



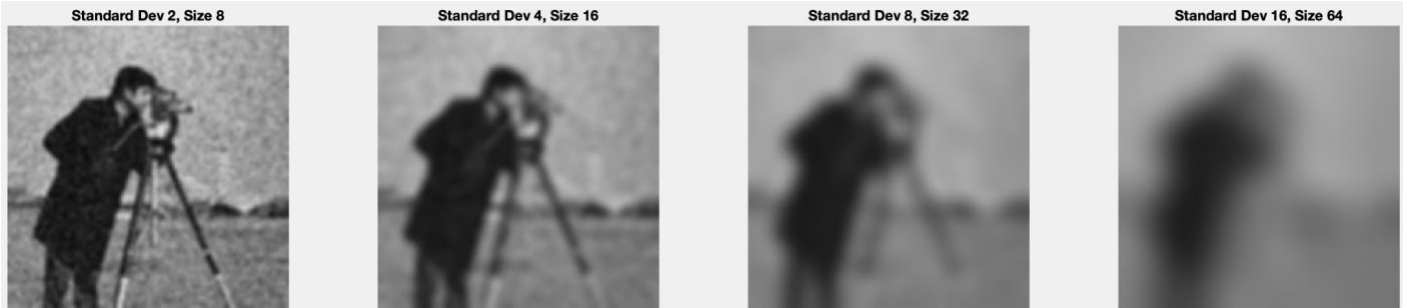
Average 8x8



Average 16x16



2. Gaussian filter of standard deviations 2, 4, 8, 16 (A good choice of gaussian filter size is 4 times its standard deviation, to include most of the variability within the box).

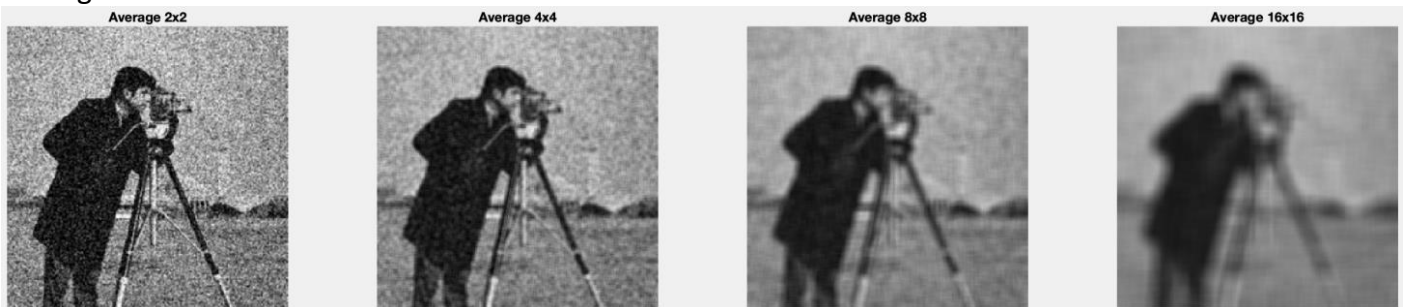


(1) Which filter works best?

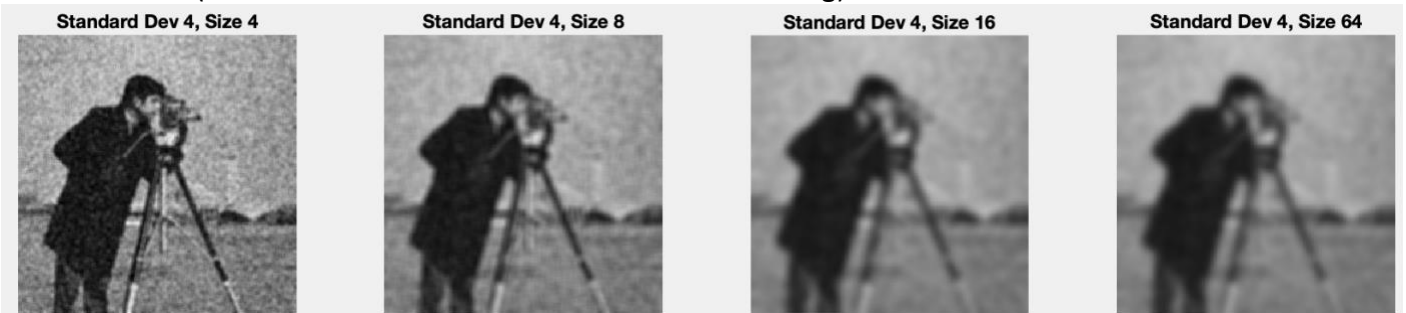
In my opinion, both filters did not perform well on this task. However, I think **the average filter works better than the Gaussian filter** since average filter still **preserves some edges while alleviated most of the salt-and-pepper problem** (like the average 4x4 filter). Gaussian filter is more like blurring both the image information (like edges) and the noise simultaneously.

(2) What happens when you vary the box filter size?

Average filter

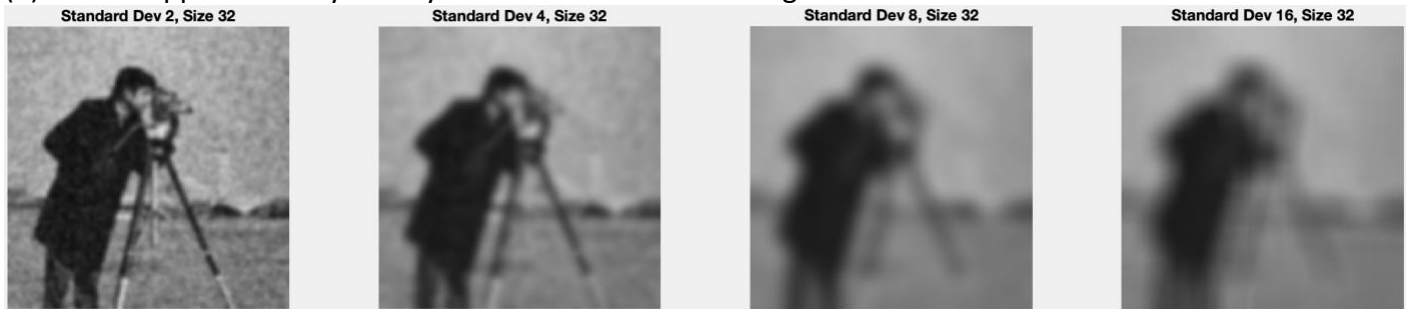


Gaussian filter (standard deviation fixed at 4 with size increasing)



From the pictures above, we can know that **both the Average filter and Gaussian filter will become blurrier if the box filter size increases.**

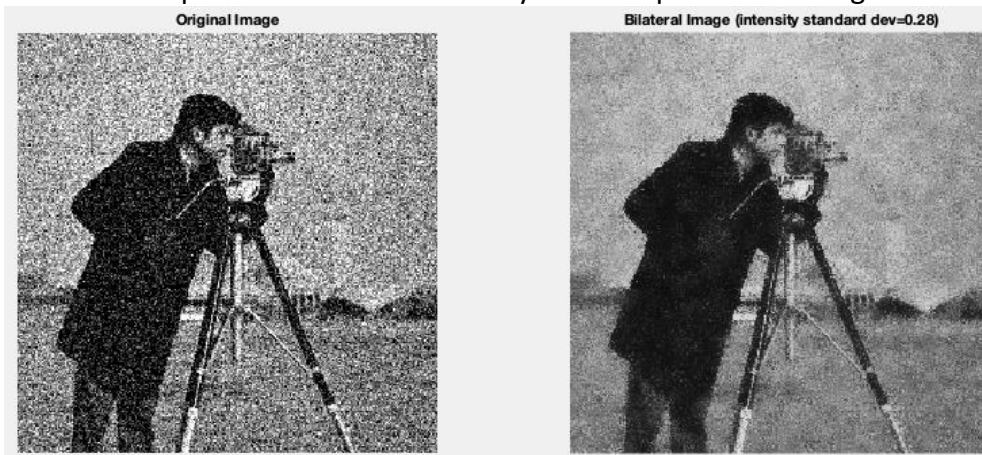
(3) What happens when you vary the standard deviation of gaussian filter?



Increasing the standard deviation of the Gaussian filter will make the picture **becomes blurrier**. Moreover, compared the picture above with the picture in the previous question that fixed the standard deviation but increased the size, we can find out that **fixing the size and increasing the standard deviation will make the picture blurry more dramatically**. Merely 4 times the standard deviation the output picture will be blurrier than 16 times the size of the box filter size.

Grad credits: Edge preserving smoothing

(1) Find the best parameters that work for you and report them along with the denoised image.

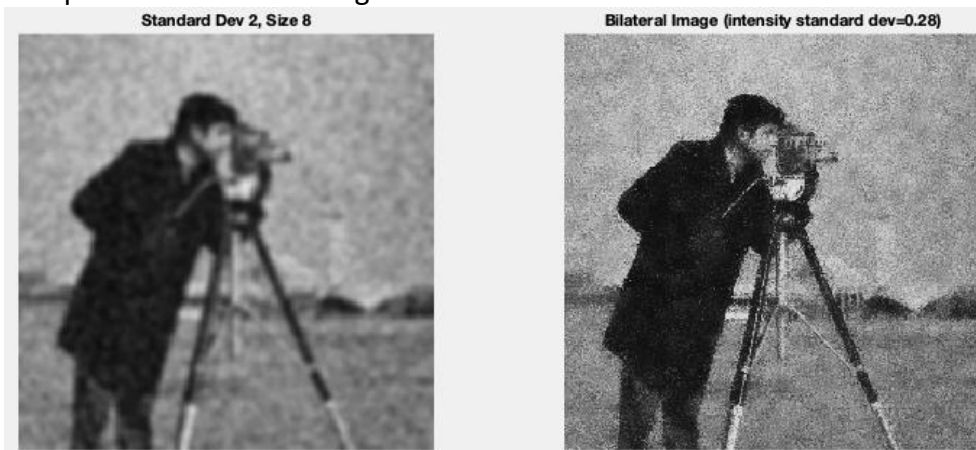


Weight = 5

Spatial standard deviation = 5

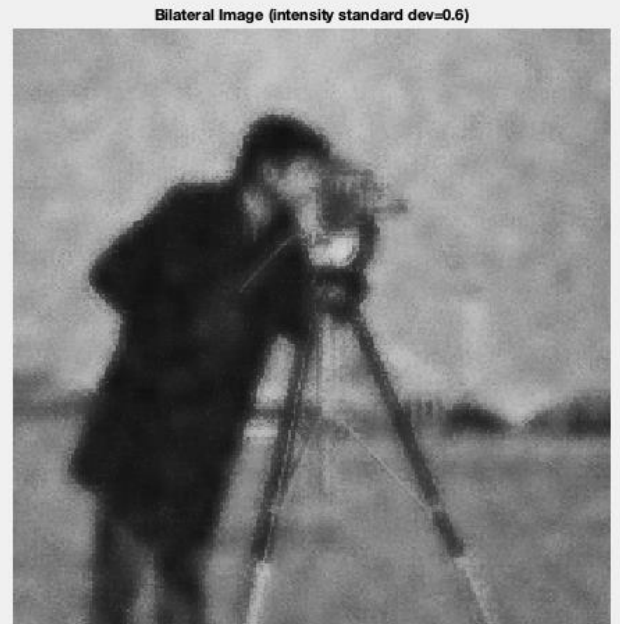
Intensity standard deviation = 0.28

(2) Compare the denoised image from bilateral filter with the one from the best Gaussian filter.



We can clearly see that the result after **Bilateral filtering** outperforms the picture after going through the best Gaussian filter with less noisy points and less blur.

(3) What will happen if we choose intensity standard deviation (σ) to be very large?



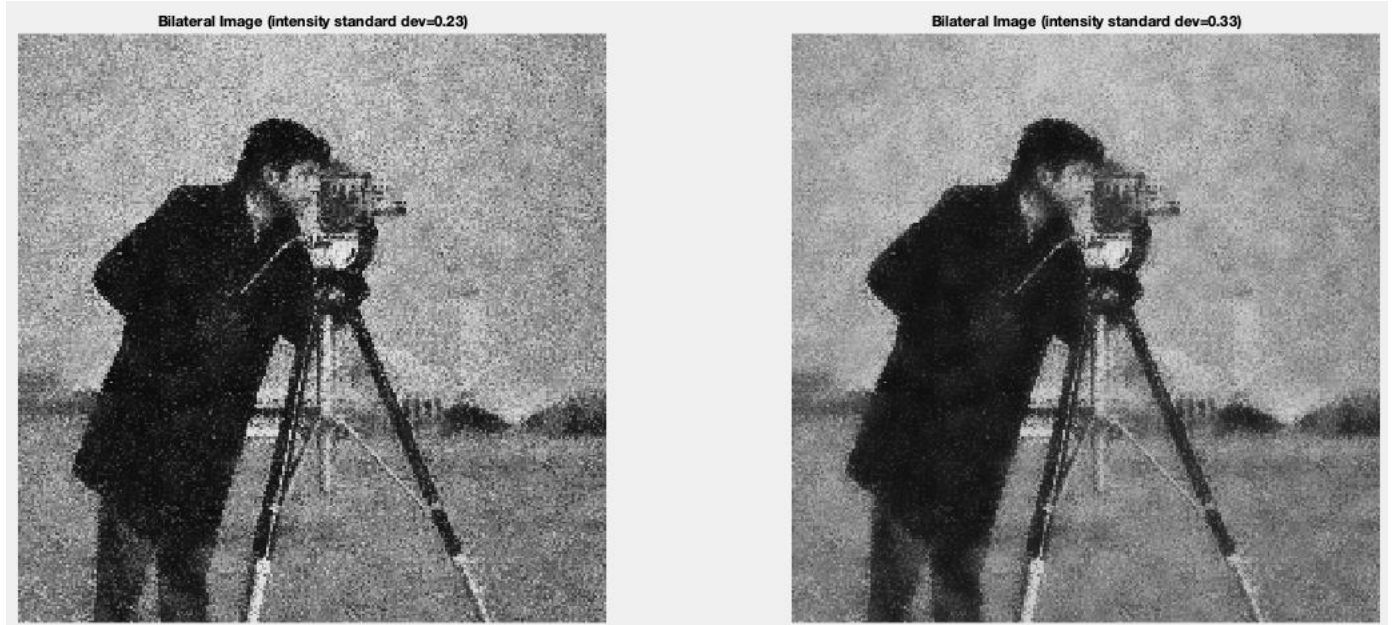
If the intensity standard deviation (σ) is larger, the filter is **more powerful at alleviating the noisy points. However, the output image is blurrier.**

(4) What will happen if we choose intensity standard deviation (σ) to be very small, say almost zero?



The output image will likely to be identical to the original image.

(5) For a particular noise standard deviation, what value (or range of values) of intensity standard deviation (σ) would you prescribe?



I would prescribe a **value range from 0.23~0.33** since the output image will be not that noisy and preserves most of the original image's information.

Problems encountered:

First, I think the tutorial has well-defined the basic functions we will use in this assignment, which helps me to better find out what specific function to use by checking the function on the MATLAB official website and helps me to forestall some function finding efforts.

The most challenging part for me in this assignment is to **implement the Central difference Gradient filter**. Since we cannot use `imgradient()` directly, we must find out how to combine the x filter and the y filter to present the result. **I eventually found out that utilizing the correct mathematical form of the filter makes the result seems a little bit darker**. Thus, I've tried lots of different ways to modify the operations and functions and eventually found out that **changing the \wedge operation to the \cdot operation makes the output more similar to the result coming from `imgradient()`**.

Also, the piecewise element calculation in MATLAB causes me some time to find out the bug of getting the wrong value since it is totally different to operate calculation element by element and to calculate it with matrix calculation.

Surprises:

I am amazed by the robustness of the bilateral filter. It is super powerful in denoising images but preserves most of the edge details. Moreover, I considered that the **Gaussian filter will be also useful if we want to hide some important information in some specific area**. Lastly, I am impressed by the **Sobel filter which efficiently highlighted the edges of the picture perfectly**.