

# CP467: Assignment 1 Report

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# Overview

- This document report the process and approaches of each question in the assignment along with output files and plot
- In the .zip file that I submitted will also include:
  - Input image : provided along with the assignment
  - Output image: the images that have been processed after the assignment
  - Main.ipynb file: this file includes all the source code to this assignment. With detailed explanation and well written comment.

## I. Image interpolation

### 1. Using the “cameraman” image as input, rescale the image to $\frac{1}{4}$ of its original size $\frac{1}{2}$ in each dimension)

- **Brief theoretical:**
  - I resize the image down to  $\frac{1}{4}$  size as its original images. First of all, I reduce the height and width of the image to  $\frac{1}{2}$  and create another empty array that have new height and width.
  - Step 2 involves copying pixels from original images to new image, I apply the SCALE of pixels to copy value from original image to new, which is skipping one pixels from original images
- **Quality result:**
  - The quality of the image has been drastically degree since the pixels size are decrease by two
  - Image looks like it has more square pixels

### 2. Using the “cameraman\_rescaled” image as input, rescale the image 4 times of its size, i.e , back to the original size of “cameraman” using the

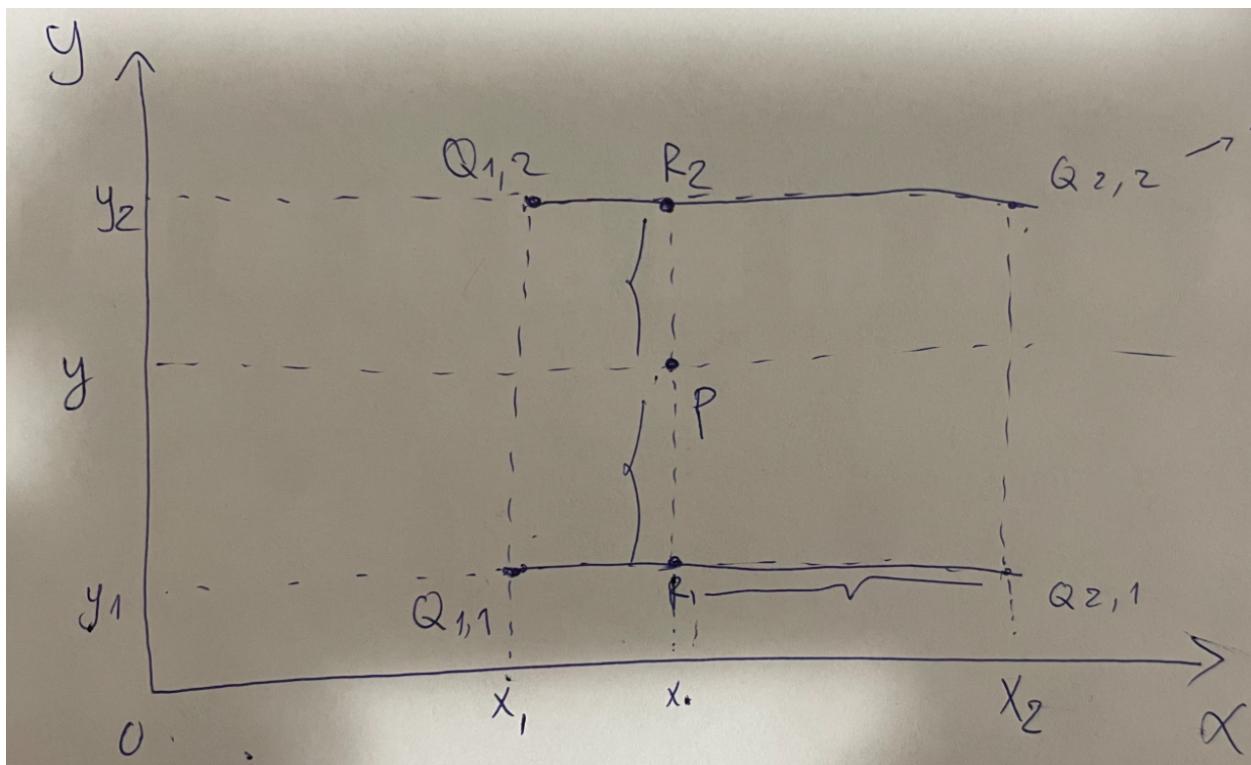
#### a. Nearest neighbor interpolation

- **Brief theoretical:**
  - The concept include identify the closest pixels and assign that value to the new pixels
  - Example:
  - [ 0 1 ] → [ 0 0 1 1 ]
  - [ 2 3 ] [ 0 0 1 1 ]
  - [ 2 2 3 3 ]
  - [ 2 2 3 3 ]
- **Quality result:**
  - Image quality has been enhanced, yet the images still remain fairly unclear, exhibiting numerous pixelated squares
  - Differences between objects has been a little more clear if we looks from far away (5 meters)

### b. Bilinear interpolation

- **Brief theoretical:**

- Key approach was to calculate the value at pixel  $(x,y)$  by identify the 4 nearest points and apply bilinear interpolation formulas
- 
- Example: matrix =
  - $[1 \_ \_ 2]$
  - $[\_ \_ \_ \_]$
  - $[\_ \_ \_ \_]$
  - $[3 \_ \_ 4]$
- Then as I draw below, we will calculate  $R_2$  and  $R_1$  using linear interpolation, after that we can calculate the value at point  $(x,y)$ . For this , we apply the same concept into bilinear interpolation



- **Quality result:**

- Using bilinear interpolation has resulted in a smoother image. However, it remains somewhat blurry

### c. Bicubic interpolation

- **Brief theoretical:**

- The bicubic interpolation is perform based on 16 nearest point and apply complex formula into the image

- **Quality result:**

- The output image generated by bicubic interpolation seems to be smoother, the edge was well formed and details are more specific, while it significant reduce the blur levels

#### d. FINAL RESULT COMPARISON



#### e. BONUS MARK : compare each image interpolation output to the original, using Mean Squared Error metrics

```
0.0s
nearestneighbor MSE: 78.387451171875
bilinear MSE: 70.14463806152344
biLinearInterpolation MSE: 100.36189270019531
```

- After applying the Mean Squared Error for 3 of the interpolation, it is interesting that bilinear interpolation has lower value than the nearestNeighbor interpolation. Does this mean the bilinear interpolation method produces worse quality image than nearestNeighbor?

- After doing some research, this is the answer that I found. MSE is used to compare the output image and the approximate transformed image compared to the original. So bilinear interpolation might seem smoother but the calculated value is lower, also there are some cons of these two methods:
  - Nearest Neighbor: produce blocky artifacts and not suitable for high quality image scaling
  - Bilinear interpolation: produce more smoother but may blur and sharp edges

## II. Point operations

### 1. Find the negative of the image

- Brief overview:**

- Negative images are simple, it is based on the rule that we're trying to find the opposite value of a pixel value. It takes the highest value in this case is 255 and minus the value at that pixels
- (this is from slide)

The negative of an image with gray levels in the range  $[0, L-1]$  is obtained by:

$$s = L-1-r$$

- 

- Quality result:**

- It is obvious that when we want to enhance bright detail in an image, this would be the best methods
- For this example, the assigned image could not fully illustrate how this works but the original image seems to get a huge brightness enhancement

### 2. Apply power-law transformation on the image

- Brief overview:**

- Produce images that are darker than intended
- Also, we use **gamma correction** to process this power-law response

Power-law transformations have the basic form:

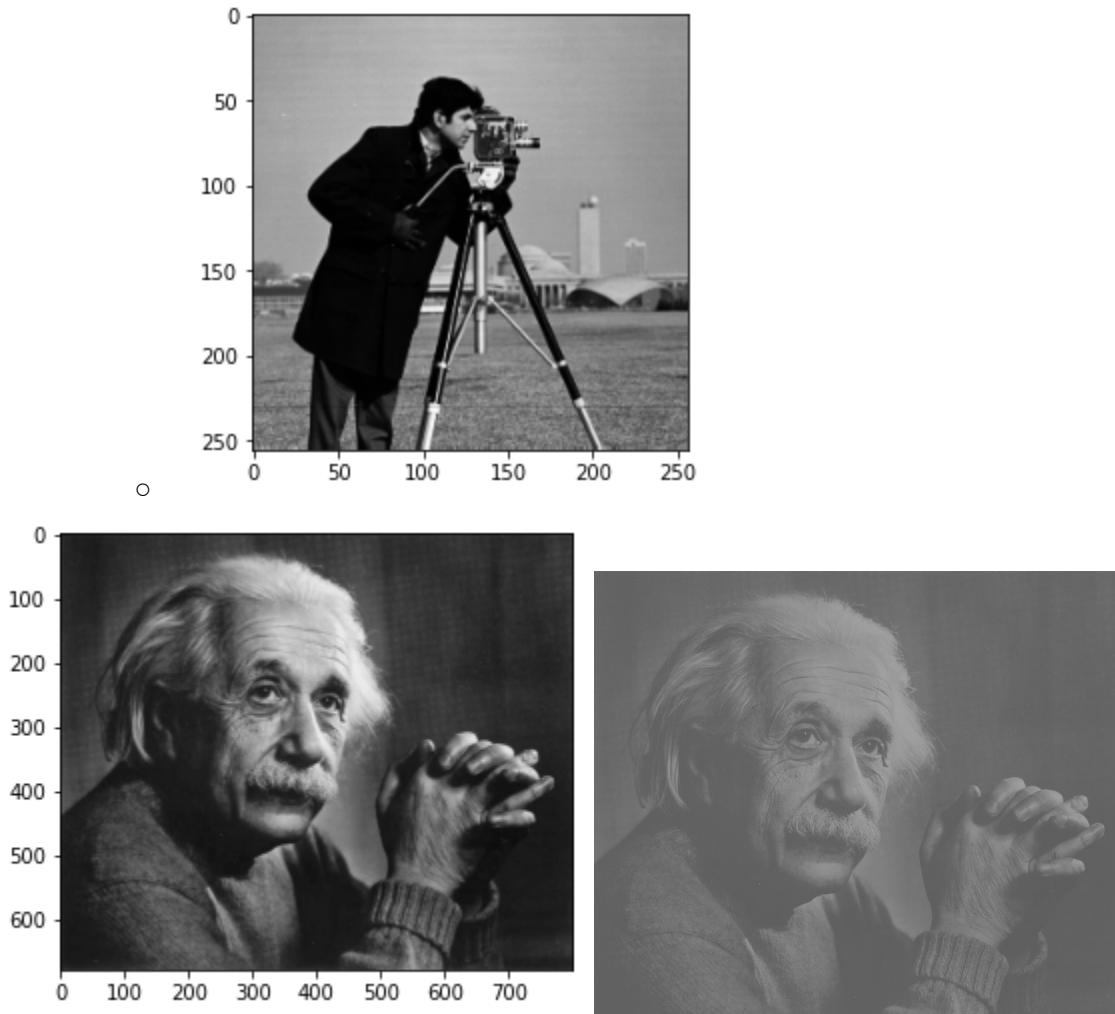
- 

$$s = cr^y$$

- The gamma value varies from 0.04, 0.10, ... 10, 25. The purpose of this is to test a wide range of values to see which will fit best with the power-law
- I chose the image with a gamma value of 0.67 because it is the clearest and we can see the cameraman's suit



- **Quality result:**
  - We can see there are many versions of the image, however based on the needs of our. We can choose which images that fits best
- 3. Apply contrast stretching on the image**
- **Brief overview:**
  - Main ideas is to increase the dynamic range of the gray level in an image
  - Just apply the formula for contrast stretching, we would have a new value that is more dynamic
- **Quality result:**
  - The quality of image has been more dynamic and looks more beautiful due to more level in gray level
  - However, using the cameraman image input would not see much different, so I do more and use Einstein instead to see the differences in more detail, this is just for illustrating the contrast stretching because Einstein input has low dynamic range of gray level



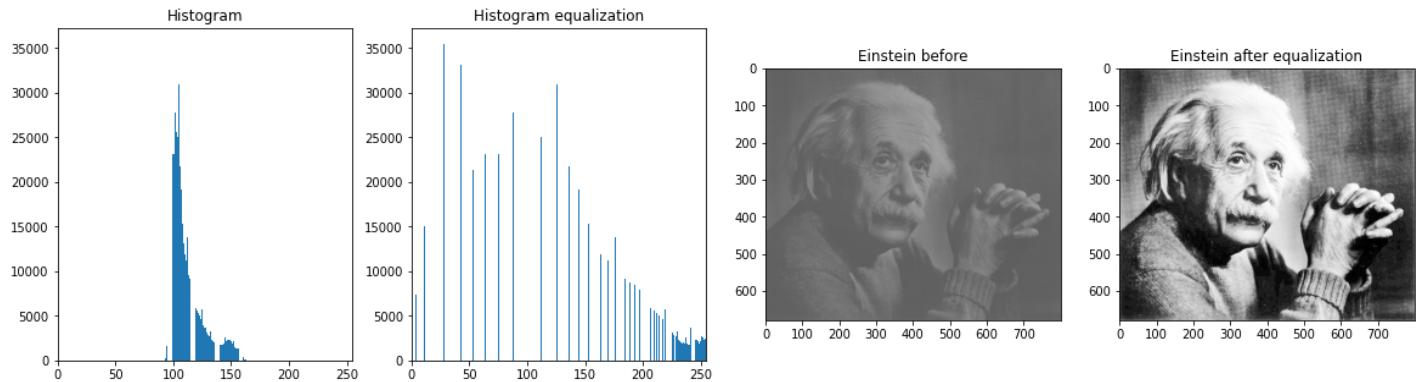
- The image on the left has been apply contrast stretching and right is the original image

### III. Histogram processing

#### 1. Histogram equalization

- **Brief overview:**

- This processing method aims to process images that are too dark or low contrast. We would try to make the value relatively equal (more equally distributed)
- For histogram equalization, the process include
  - Compute the histogram of the image (which is shown below in the image, before and after equalization)
  - Compute CDF function and then equalization if performed



- **Quality result:**

- Before: the image is dark, it is hard to witness Einstein
- After: after the image being equalize, the portrait of Einstein has been highlighted and more eye friendly

## 2. Histogram matching (specification)

- Brief overview:

- In this problem, we attempt to enhance on the images given the shape of the target image (or we compare in histogram), it is useful to know the target shape.
- At first, we all have to calculate the PDF, CDF of a target and original image, then a mapping phrase would be applied to get the final output image.

- Quality result:

- As we want to enhance the original image, it is really hard to look at the bones in the image. However, we have the histogram shape of the target that we wanted to reach, now it is easy that we can use both values and the output of the image has been almost the same to the target.

