

DAIICT

Assignment 1

Advanced Image Processing

Kishan Vaishnani - 202011004

Dhyani Mehta - 202011032

Date: 29-1-2021

Table of Contents

Task 1.....	2
Problem Statement.....	2
Library Used.....	2
Method.....	2
1. Up & Down Sampling.....	2
2. Quantization level	2
Observation	2
1. Up-Sample & Down-Sample	2
2. Effect of different quantization levels.....	3
Conclusion.....	4
Task 2.....	5
Problem Statement.....	5
Library Used.....	5
Method.....	5
1. Computing Mean.....	5
2. Computing Variance.....	5
Observation	5
1. Effect of white gaussian noise of mean 0 and standard deviation 20	5
2. Some noisy images of fig1 and average image of all of them	6
Conclusion.....	7
Task 3.....	8
Problem Statement.....	8
Library used.....	8
Method.....	8
1. Converting a grayscale image into Binary Image.....	8
2. Finding Connected Components	8
Observation	9
Conclusion.....	9

Task 1

Problem Statement

Up-sample and down-sample the image of Figure 1 by scale factor 4. Discuss the effect of changing sampling rate. Also observe the effect of different quantization levels ($L = 2, 4, 8, 16, 32, 64, 128, 256$) for this image. (Figure 1 image: fig1.jpg)

Library Used: PIL

Method

1. Up & Down Sampling

PIL.Image.Image.resize((Width * Scaling factor, Height * Scaling factor))

We have used the PIL library's resize function for sampling the image, which by default uses Bicubic Interpolation.

For up-sampling, scaling factor = 4 and,

For Down-sampling, scaling factor = $\frac{1}{4}$

2. Quantization level

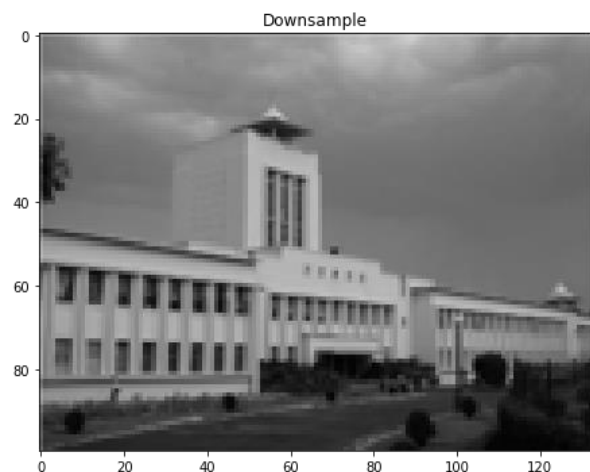
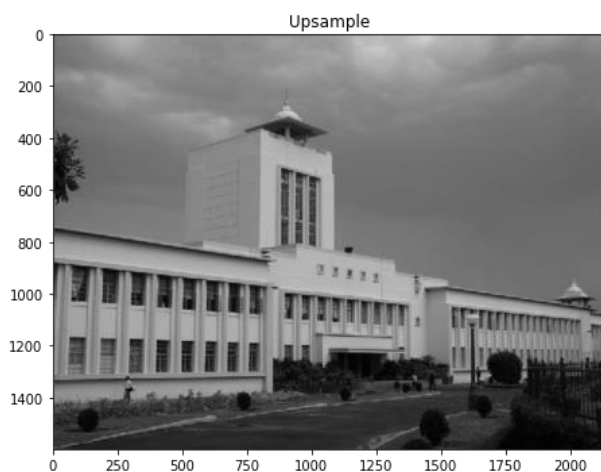
PIL.Image.Image.quantize(L)

Return new image with applied quantization level L where $L = 2^k$, $k = \text{Bits}$

Observation

1. Up-Sample & Down-Sample

Up-sample size: (2136, 1600), Down-sample size: (133, 100)



2. Effect of different quantization levels

Observation: By increasing level we have observed image appears clearer than last level.



L=2



L=4





L=8



L=16



L=32	L=64
 <p>L=128</p>	 <p>L=256</p>

Conclusion

- Up-sampling the image increases its size by the sampling factor, and the image becomes zoomed, but the quality is reduced.
- Down-sampling decreases its size by the sampling factor and the image becomes more blurred, and quality is significantly reduced.

Task 2

Problem Statement

Compute the mean and variance of Figure 2. Compute the same parameters for Figure 1. Then add additive white Gaussian noise of mean 0 and standard deviation 20 to both the images and compute the means and variances. Explain your observations. Further, generate several such noisy images from Figure 1 and average all of them. Does the averaged image have a less noisy artifact? (Figure 2 image: baboon.jpg)

Library Used: NumPy

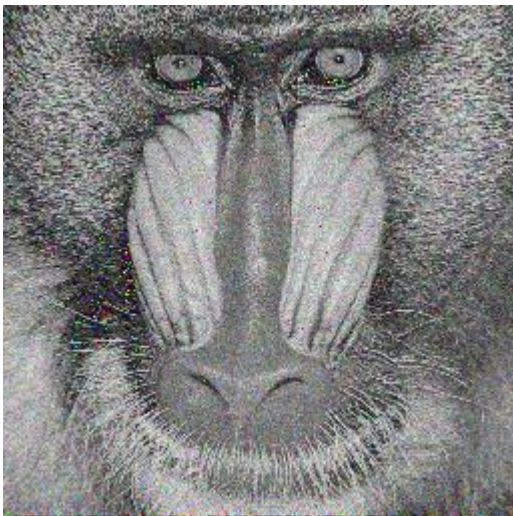
Method

1. Computing Mean
`numpy.mean(image_array)`
2. Computing Variance
`numpy.var(image_array)`

Here PIL Image object is converted to a NumPy array, and then the mean and variance are computed.

Observation

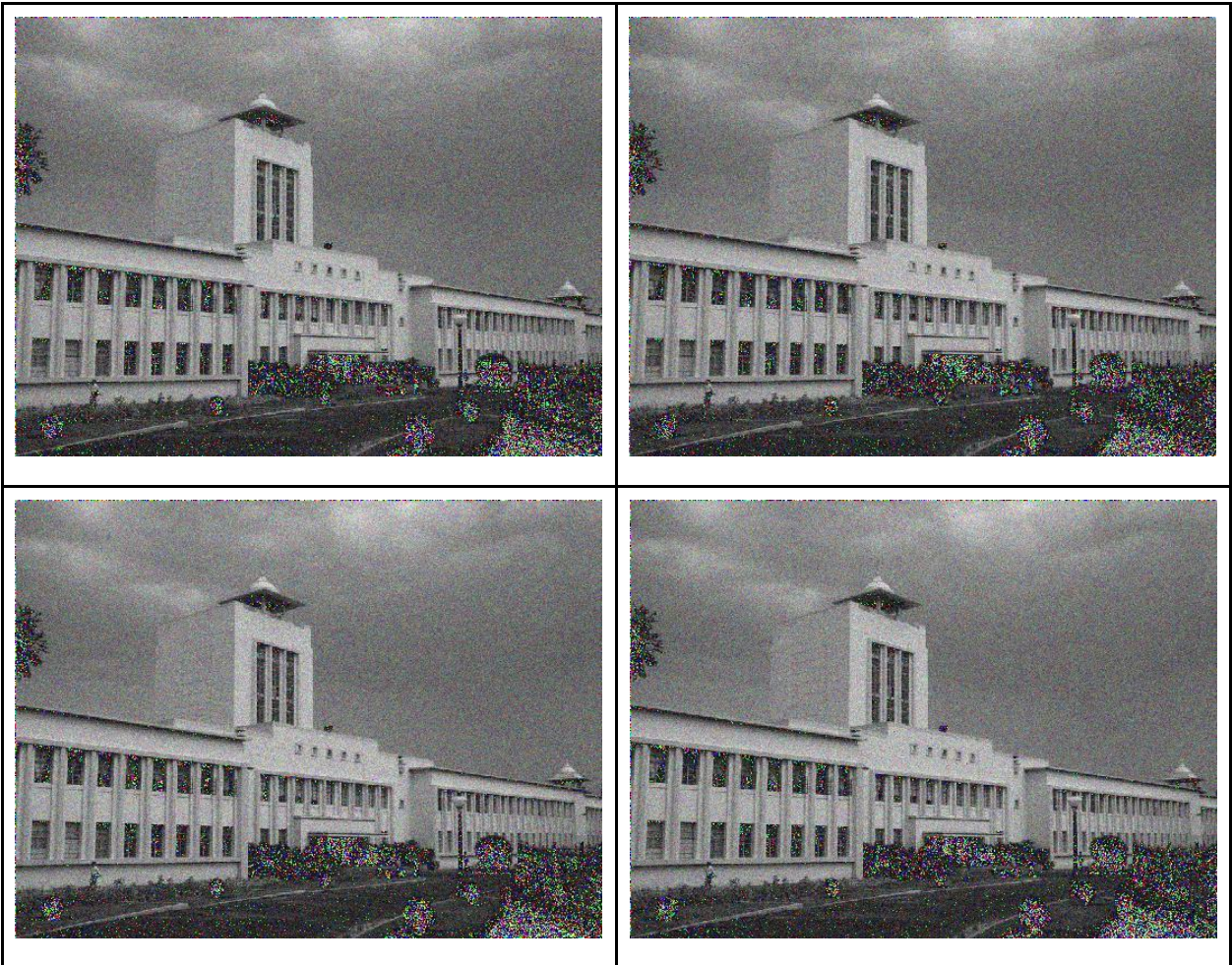
1. Effect of white gaussian noise of mean 0 and standard deviation 20
Observation: By adding noise to an image, it appears with white and random colored spots and gets little bit unclear. Also, their mean and variance gets increased.

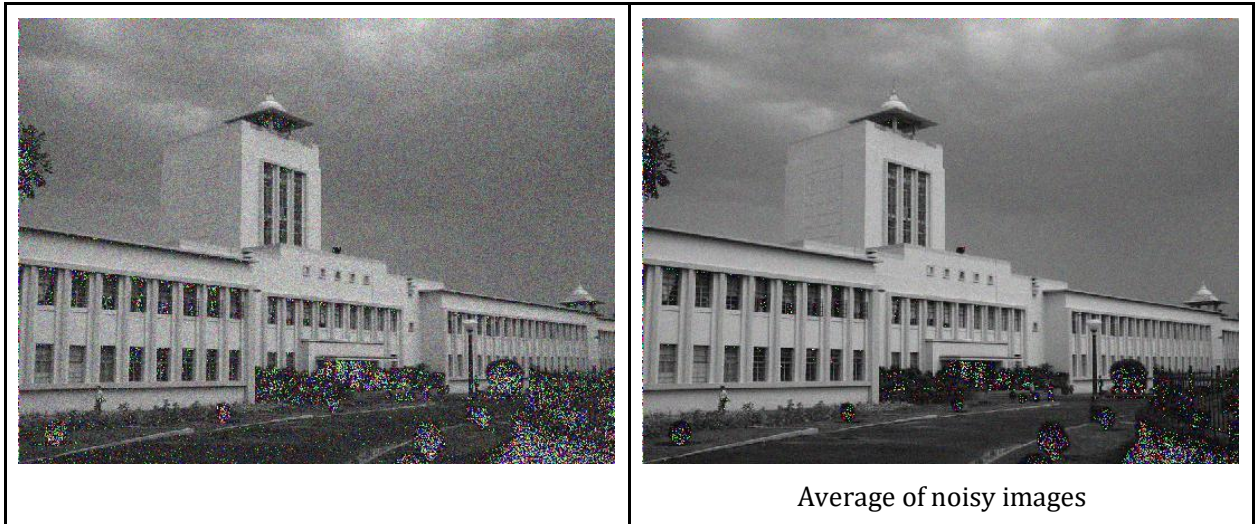


Without noise: Mean: 128.96, Variance: 1820.05	Without noise: Mean: 112.94, Variance: 2133.13
With noise: Mean: 129.12, Variance: 2206.51	With noise: Mean: 116.15, Variance: 2512.29

2. Some noisy images of fig1 and average image of all of them

Observation: Averaging more noisy images gives an image with even less noisy artifacts.





Conclusion

- As the quantization level increases, the image can use more values to display the image, and thus more and more fine details can be seen.
- The mean and variance of the image increases as the image becomes noisier as the noise are merely random values.
- The image quality is improved by averaging its noisy images together.

Task 3

Problem Statement

Find out the number of objects from Figure 3. Label distinct objects with distinct colors.
[Use the algorithm of finding out connected components.] (Figure 3 image: cc.jpg)

Library used: OpenCV

Method

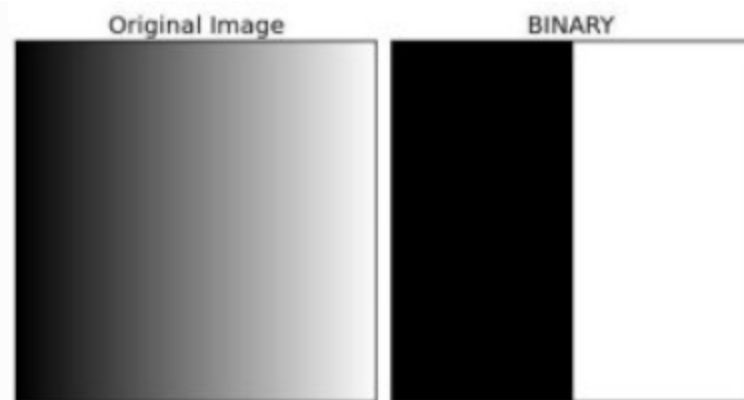
1. Converting a grayscale image into Binary Image

cv2.threshold(a,b,c,d)

- a. Greyscale image to be convert
- b. A threshold value is used to classify the pixel values.
- c. maxVal will assign to a pixel if the pixel value is more/less than the threshold value.
- d. Different types of thresholding function.

cv2.THRESH_BINARY

This thresholding function will specify the threshold to be converted into pixel values.



Source: <https://opencv-python-tutroals.readthedocs.io/>

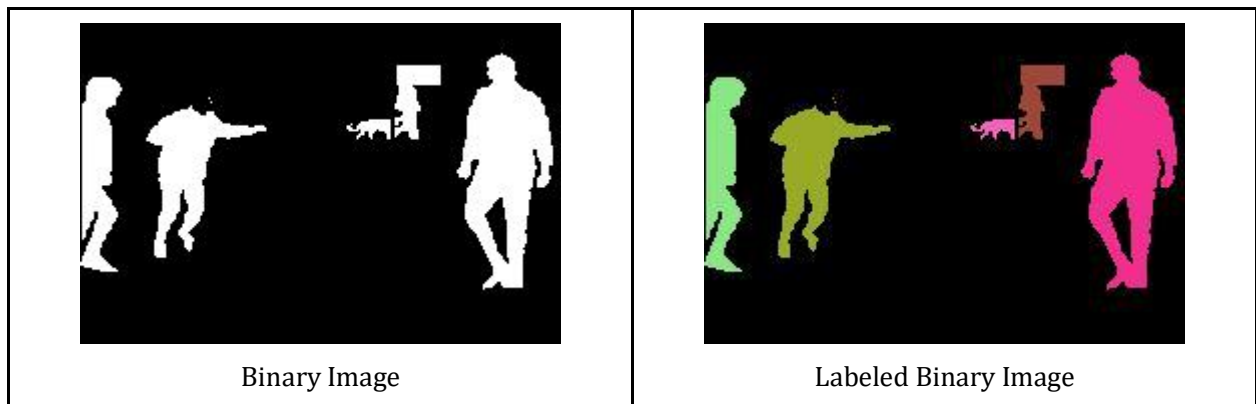
2. Finding Connected Components

cv2.connectedComponentsWithStats(BinaryImage, Connectivity)

This method will return following Parameters.

- a. Total number of labels it had generated
- b. A labeled set similar size to Binary Image

Observation



Conclusion

- By converting the image to a binary image (i.e., $L = 2$), we can easily find connected components in the image.