```
In [2]: original_image = cv2.imread('foundationscomputerscience.jpg', 0)
         plt.imshow(original_image)
Out[2]: <matplotlib.image.AxesImage at 0x7fbc31767410>
          100
          200
          300
          400
          500
          600
          700
          800
                           600
                                800
                                     1000 1200 1400
In [3]: image_pixels = np.array([original_image])
         image_pixels
Out[3]: array([[[25, 22, 26, ..., 16, 16, 16],
                 [27, 24, 26, ..., 16, 16, 16],
                 [27, 24, 23, ..., 16, 16, 16],
                 [21, 29, 22, ..., 16, 16, 16],
                 [27, 22, 19, ..., 16, 16, 16],
                 [76, 52, 46, ..., 16, 16, 16]]], dtype=uint8)
In [4]: plt.hist(image_pixels.ravel(), 256, [0,256])
         plt.show()
          80000
          70000
          60000
          50000
          40000
          30000
          20000
          10000
                              100
                                     150
                                                   250
 In [ ]: #image_1 = original_image
         #plt.imshow(image_1, cmap='RdY1Gn')
 In [6]: def show(image):
           plt.imshow(image, cmap = 'gray')
           plt.xticks([])
           plt.yticks([])
 In [7]: height, width = original_image.shape
         img = np.zeros((height + 160, width), np.uint8)
         img[80:-80, :] = original_image
         plt.figure(figsize=(15,5))
         plt.subplot(131)
         show(img)
 In [8]: gray_pixels = np.array([img])
         gray_pixels
Out[8]: array([[[0, 0, 0, ..., 0, 0, 0],
                 [0, 0, 0, ..., 0, 0, 0],
                 [0, 0, 0, \ldots, 0, 0, 0],
                 [0, 0, 0, \ldots, 0, 0, 0],
                 [0, 0, 0, ..., 0, 0, 0],
                 [0, 0, 0, ..., 0, 0, 0]]], dtype=uint8)
 In [9]: plt.hist(img.ravel(), 256, [0,256])
         plt.show()
          250000
          200000
          150000
          100000
           50000
                                     150
                                             200
                                                    250
In [10]: !pip install sewar
         Requirement already satisfied: sewar in /usr/local/lib/python3.7/dist-packages (0.4.4)
         Requirement already satisfied: Pillow in /usr/local/lib/python3.7/dist-packages (from sewar) (7.1.2)
         Requirement already satisfied: scipy in /usr/local/lib/python3.7/dist-packages (from sewar) (1.4.1)
         Requirement already satisfied: numpy in /usr/local/lib/python3.7/dist-packages (from sewar) (1.21.5)
In [11]: from sewar import full_ref
         from skimage import measure, metrics
         from PIL import Image
         from google.colab.patches import cv2_imshow
         Applying the mean filter with 3x3 kernel
In [12]: mean_kernel = np.ones((3, 3), np.float32) / 25
In [13]: convolution_mean = cv2.filter2D(img, -1, mean_kernel, borderType = cv2.BORDER_CONSTANT)
         plt.figure(figsize=(15, 5))
         show(convolution_mean)
In [14]: rmse_skimg = metrics.normalized_root_mse(img, convolution_mean)
         print('RMSE: based on scikit-image = ', rmse_skimg)
         RMSE: based on scikit-image = 0.6512582223038206
In [15]: mse_skimg = metrics.mean_squared_error(img, convolution_mean)
         print('MSE: based on scikit-image = ', mse_skimg)
         MSE: based on scikit-image = 1586.5428254881372
In [16]: psnr_skimg = metrics.peak_signal_noise_ratio(img, convolution_mean, data_range=None)
         print('PSNR: based on scikit-image = ', psnr_skimg)
         PSNR: based on scikit-image = 16.126285613793765
In [17]: from skimage.metrics import structural_similarity as ssim
         ssim_skimg = ssim(img, convolution_mean, data_range = img.max() - img.min(), multichannel = True)
         print('SSIM: based on scikit-image = ', ssim_skimg)
         SSIM: based on scikit-image = 0.599734707888257
         Applying the mean filter with 5x5 kernel
In [18]: mean_kernel = np.ones((5, 5), np.float32) / 25
In [19]: convolution_mean = cv2.filter2D(img, -1, mean_kernel, borderType = cv2.BORDER_CONSTANT)
         plt.figure(figsize=(15, 5))
         show(convolution_mean)
In [20]: rmse_skimg = metrics.normalized_root_mse(img, convolution_mean)
         print('RMSE: based on scikit-image = ', rmse_skimg)
         RMSE: based on scikit-image = 0.19344577109999714
In [21]: mse_skimg = metrics.mean_squared_error(img, convolution_mean)
         print('MSE: based on scikit-image = ', mse_skimg)
         MSE: based on scikit-image = 139.97930755299882
In [22]: psnr_skimg = metrics.peak_signal_noise_ratio(img, convolution_mean, data_range=None)
         print('PSNR: based on scikit-image = ', psnr_skimg)
         PSNR: based on scikit-image = 26.670165200449606
In [23]: from skimage.metrics import structural_similarity as ssim
         ssim_skimg = ssim(img, convolution_mean, data_range = img.max() - img.min(), multichannel = True)
         print('SSIM: based on scikit-image = ', ssim_skimg)
         SSIM: based on scikit-image = 0.8875961528181259
In [23]:
         Applying the mean filter with 8x8 kernel
In [24]: mean_kernel = np.ones((8, 8), np.float32) / 25
In [25]: convolution_mean = cv2.filter2D(img, -1, mean_kernel, borderType = cv2.BORDER_CONSTANT)
         plt.figure(figsize=(15, 5))
         show(convolution_mean)
In [26]: rmse_skimg = metrics.normalized_root_mse(img, convolution_mean)
         print('RMSE: based on scikit-image = ', rmse_skimg)
         RMSE: based on scikit-image = 1.3667322308384877
In [27]: | mse_skimg = metrics.mean_squared_error(img, convolution_mean)
         print('MSE: based on scikit-image = ', mse_skimg)
         MSE: based on scikit-image = 6987.34574137485
In [28]: psnr_skimg = metrics.peak_signal_noise_ratio(img, convolution_mean, data_range=None)
         print('PSNR: based on scikit-image = ', psnr_skimg)
         PSNR: based on scikit-image = 9.687681277242984
In [29]: from skimage.metrics import structural_similarity as ssim
         ssim_skimg = ssim(img, convolution_mean, data_range = img.max() - img.min(), multichannel = True)
         print('SSIM: based on scikit-image = ', ssim_skimg)
         SSIM: based on scikit-image = 0.5573752691359672
In [29]:
         Applying the mean filter with 10x10 kernel
In [30]: mean_kernel = np.ones((10, 10), np.float32) / 25
In [31]: convolution_mean = cv2.filter2D(img, -1, mean_kernel, borderType = cv2.BORDER_CONSTANT)
         plt.figure(figsize=(15, 5))
         show(convolution_mean)
In [32]: rmse_skimg = metrics.normalized_root_mse(img, convolution_mean)
         print('RMSE: based on scikit-image = ', rmse_skimg)
         RMSE: based on scikit-image = 2.060715437808864
In [33]: | mse_skimg = metrics.mean_squared_error(img, convolution_mean)
         print('MSE: based on scikit-image = ', mse_skimg)
         MSE: based on scikit-image = 15884.78751769164
In [34]: | psnr_skimg = metrics.peak_signal_noise_ratio(img, convolution_mean, data_range=None)
         print('PSNR: based on scikit-image = ', psnr_skimg)
         PSNR: based on scikit-image = 6.120989509909393
In [35]: from skimage.metrics import structural_similarity as ssim
         ssim_skimg = ssim(img, convolution_mean, data_range = img.max() - img.min(), multichannel = True)
         print('SSIM: based on scikit-image = ', ssim_skimg)
         SSIM: based on scikit-image = 0.40245689065915996
```

In []: !jupyter nbconvert --to html Seatwork_1.ipynb

In [1]: import numpy as np

import cv2

from matplotlib import pyplot as plt