Part 1

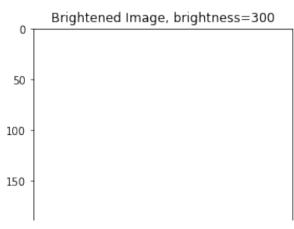
Notes

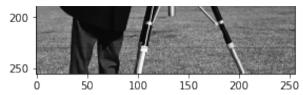
imread is deprecated in SciPy 1.0.0, and will be removed in 1.2.0. So I use imageio.imread in the following instead.

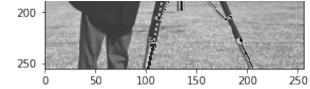
```
# Read the image
 8 | cameraman origin = imageio.imread('cameraman.tif')
  # Create a copy of the origina image for us to manipulate
   cameraman bright 50 = copy.deepcopy(cameraman origin)
   cameraman bright 300 = copy.deepcopy(cameraman origin)
12
13
   # Call imadd to perform enhancement
   cameraman bright 50 = imadd(cameraman bright 50.50)
   cameraman bright 300 = imadd(cameraman bright 300,300)
15
16
   # Show the results
17
18 plt.figure(figsize=(15,6))
19 plt.subplot(131)
20 plt.title('Original Image')
21 plt.imshow(cameraman origin,cmap='gray',vmin = 0, vmax = 255)
22 plt.subplot(132)
23 plt.title('Brightened Image, brightness=50')
24 plt.imshow(cameraman bright 50, cmap='gray', vmin = 0, vmax = 255)
25 plt.subplot(133)
26 plt.title('Brightened Image, brightness=300')
   plt.imshow(cameraman bright 300,cmap='gray',vmin = 0, vmax = 255)
28 plt.show()
```

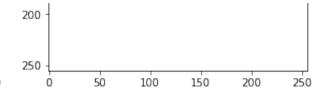












In [4]: numbers of distinct pixel values in the above three images are respectively', ip.unique(cameraman_origin)), len(np.unique(cameraman_bright_50)), len(np.unique(cameraman_bright_300))

The numbers of distinct pixel values in the above three images are respectively 247 247 247

Comments

The dynamic range (the number of distinct pixel values in an image) of the original and the enhanced image are both 247, because simply adding a constant does not affect the number of distinct values.

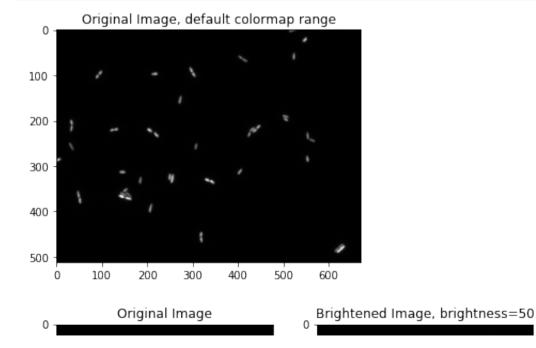
But since the gray-scale value range of uint8 is 0-255, the pixels whose value exceeds 255 will appear white, hence the number of pixels with distinct color may decrease when we brighten an image.

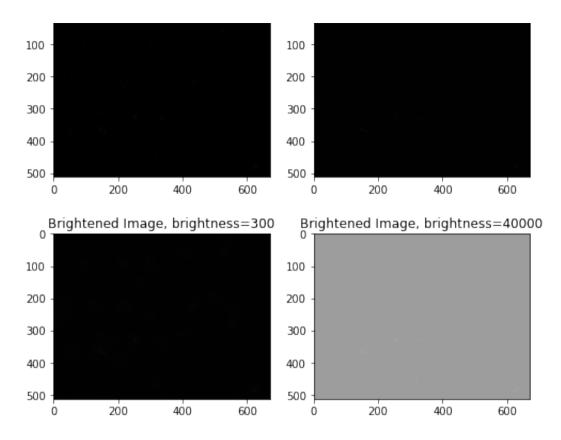
When we increase the brightness by 300, all pixel values will exceeds 255, so the picture looks purely white.

```
In [5]: 1 # Read the image
2 eco_origin = imageio.imread('eco.tif')
3 # Create a copy of the origina image for us to manipulate
4 eco_bright_50 = copy.deepcopy(eco_origin)
5 eco_bright_300 = copy.deepcopy(eco_origin)
6 eco_bright_40000 = copy.deepcopy(eco_origin)
7

8 # Call imadd to perform enhancement
9 eco_bright_50 = imadd(eco_bright_50,50)
10 eco_bright_300 = imadd(eco_bright_300,300)
11 eco_bright_40000 = imadd(eco_bright_40000,40000)
12
13 # Show the results
```

```
14 | plt.figure()
15 plt.title('Original Image, default colormap range')
16 plt.imshow(eco origin.cmap='grav')
17 plt.figure(figsize=(8.7))
18 plt.subplot(221)
19 plt.title('Original Image')
20 plt.imshow(eco_origin,cmap='gray', vmin=0, vmax=2**16-1)
21 plt.subplot(222)
22 plt.title('Brightened Image, brightness=50')
23 plt.imshow(eco bright 50,cmap='gray', vmin=0, vmax=2**16-1)
24 plt.subplot(223)
25 plt.title('Brightened Image, brightness=300')
26 plt.imshow(eco bright 300,cmap='gray', vmin=0, vmax=2**16-1)
   plt.subplot(224)
28 plt title('Brightened Image, brightness=40000')
29 plt.imshow(eco bright 40000,cmap='gray', vmin=0, vmax=2**16-1)
30 plt.show()
```





The numbers of distinct pixel values in the above three images are respectively 1748 1748 1748

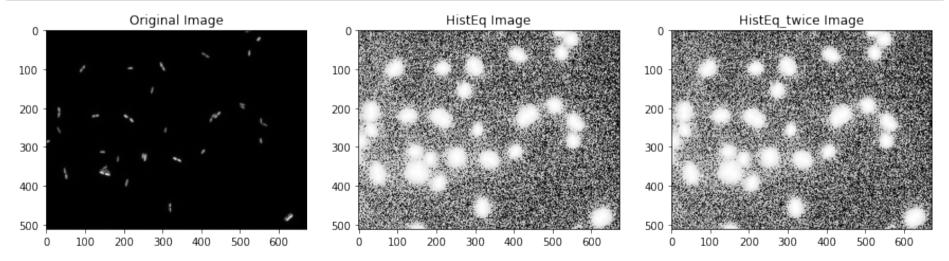
The quality of the image eco.tif cannot be enhanced by simply increasing its brightness.

Because the gray-scale value range of uint16 is 0-65535, the brightness increment 50 and 300 are too small compared to the range, so the image looks completely black. If we increase the brightness to a larger value like 40000, it still looks completely grey.

That's because, as we can see from the image with default colormap range (in which the colormap covers the complete value range of the supplied data), the gray scale values of the eco.tif is so unevenly distributed that most of the pixels are dark while the bright pixels are sparse. So adding a constant to every pixels will not increase the quality of the image.

Part 2

```
In [7]:
            # Read the image
            eco_origin = imageio.imread('eco.tif')
            # Apply Histogram Equalization here!
            eco_histeg = exposure.equalize_hist(eco_origin)
            eco histeg twice = exposure.equalize hist(eco histeg)
            # Show the results
            plt.figure(figsize=(15,6))
           plt.subplot(131)
            plt.title('Original Image')
            plt.imshow(eco_origin,cmap='gray')
        13 plt.subplot(132)
        14 plt.title('HistEq Image')
            plt.imshow(eco_histeq,cmap='gray')
        16 plt.subplot(133)
        17 plt.title('HistEq_twice Image')
        18 plt.imshow(eco_histeq_twice,cmap='gray')
            plt.show()
```



Question

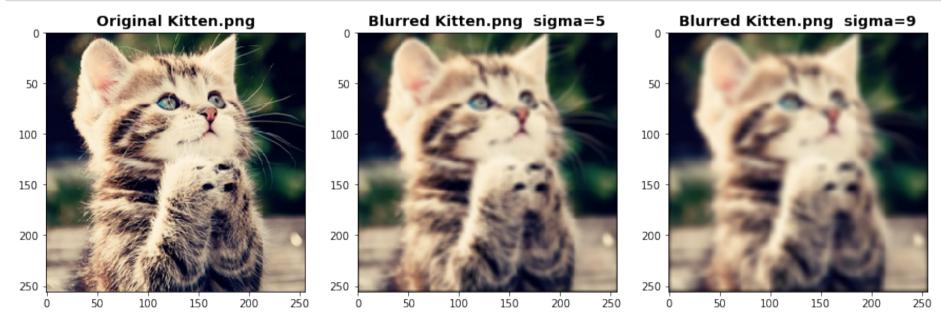
Q: Can you improve the result of enhancement by repeating the histogram equalization? Why?

• No. If an image is histogram equalized, its gray values are uniformly distributed, repeating the histogram equalization on it will just produce the same result.

Part 3

```
In [8]:
            # Gaussian Kernel Following the Descriptiong:
            # http://www.mathworks.com/help/images/ref/fspecial.html
            def gengaussian(size=5.sigma=3.0);
                 if size%2==0 or size<2:</pre>
                     print('Size Not Valid')
                     return None
                 kernel = np.zeros((size.size))
                 for x in range(size):
          8
          9
                     for y in range(size):
                         kernel[x][y] = np.exp(-((x-(size-1)/2)**2)
         10
         11
                                        +(y-(size-1)/2)**2)/(2*sigma**2))
         12
                 kernel = kernel / np.sum(kernel)
         13
                 return kernel
         14
         15 | # Read Image and Display
         16 kitten origin = imageio.imread('kitten.png')
         17 # Create a copy of the origina image for us to manipulate
         18 kitten blur 5 = copy.deepcopy(kitten origin)
         19 | kitten_blur_9 = copy.deepcopy(kitten origin)
         20 # Generate Kernel
         21 kernel_5 = gengaussian(5)
            kernel_9 = gengaussian(9)
```

```
# Apply Convolution Here!
   for i in range(3):
25
       kitten blur 5[:,:,i] = signal.convolve2d(kitten origin[:,:,i], kernel 5, mode='same')
       kitten blur 9[:,:,i] = signal.convolve2d(kitten origin[:,:,i], kernel 9, mode='same')
26
27
28 # Display Results
29 plt.figure(figsize=(15,6))
30 plt.subplot(131)
31 plt.title('Original Kitten.png', fontsize=14, fontweight='bold')
32 plt.imshow(kitten origin,vmin = 0, vmax = 255)
33 plt.subplot(132)
34 plt.title('Blurred Kitten.png sigma=5', fontsize=14, fontweight='bold')
35 plt.imshow(kitten_blur_5,vmin = 0, vmax = 255)
36 plt.subplot(133)
37 plt.title('Blurred Kitten.png sigma=9', fontsize=14, fontweight='bold')
38 plt.imshow(kitten blur 9,vmin = 0, vmax = 255)
39 plt.show()
```



A larger sigma produce a more blurry image.