

# **Computer Vision (SPRING 2016) Problem Set #1**

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# 1a: Interesting Images



Image 1 - ps1-1-a-1.png

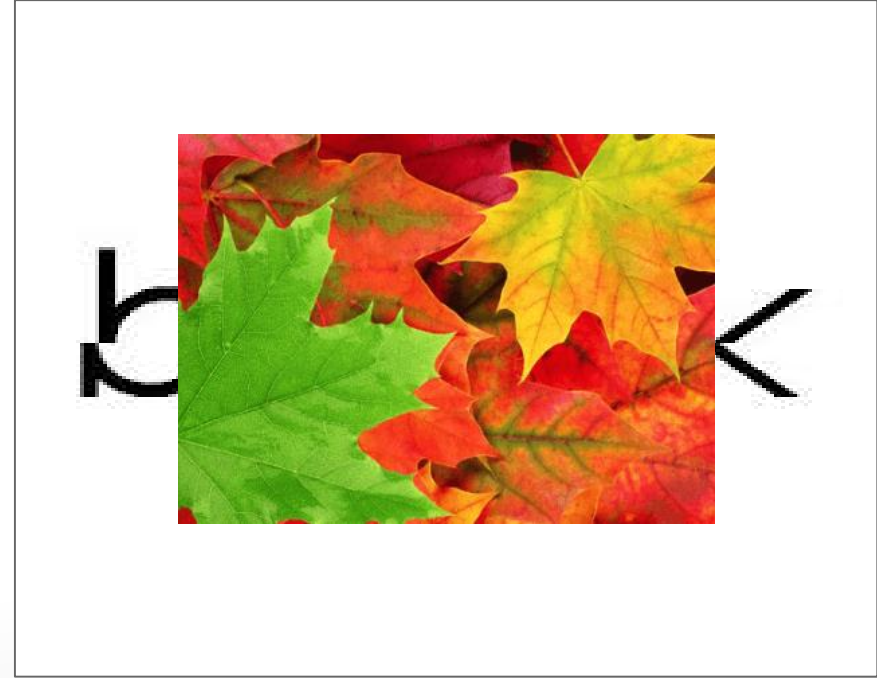


Image 2 - ps1-1-a-2.png

## 2a: Swapped Red and Blue



ps1-2-a-1.png

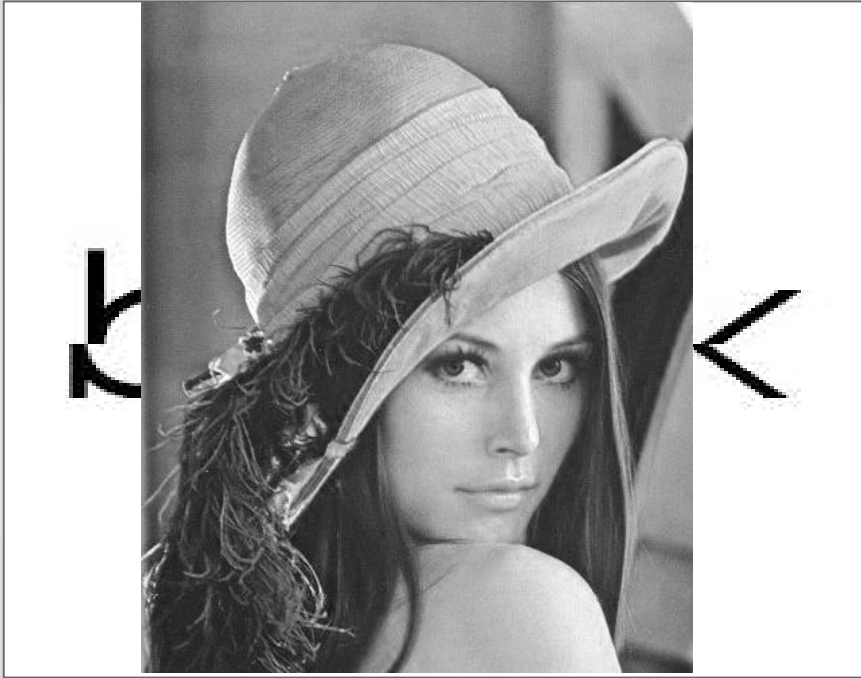
## 2b: Monochrome Green



The left image is the monochrome green image. For comparison, a grayscale image converted from the original BGR image is stored in the “temp” folder.

**Img1\_green - ps1-2-b-1.png**

## 2c: Monochrome Red



Img1\_red - ps1-2-c-1.png

## 2d: Expectations

The green monochrome image most resembles the grayscale conversion of the original image. This is the same as what I thought. Although human eyes has three kinds of cone cells which are sensitive to red, green, and blue, respectively, from luminosity point of view green colors (555 nm) contribute the most towards human perception of brightness.

When converting an image to a grayscale in openCV, the converted intensity is calculated from  $0.114 \cdot \text{Blue} + 0.587 \cdot \text{Green} + 0.299 \cdot \text{Red}$ . We see the grayscale image has the most “green” portion.

To simulate human vision system, I expect CV algorithm to work better on the green color. Depending on different applications like military, weather, etc., CV algorithm also expect to work better on other wavelengths like GHz, THz, infrared, visible, and UV.

# 3a: Replacement of Pixels



**ps1-3-a-1.png**

## 4a: Geometric Values

The standard deviation of img1\_green is 41.0

The mean of img1\_green is 78.0

##Code

```
import numpy as np
standard_deviation = np.round(np.std(img1_green))
mean_value = np.round(np.mean(img1_green))
print ('The standard deviation of img1_green is',
standard_deviation)
print ('The mean of img1_green is', mean_value)
```



# 4b: Arithmetic Operation



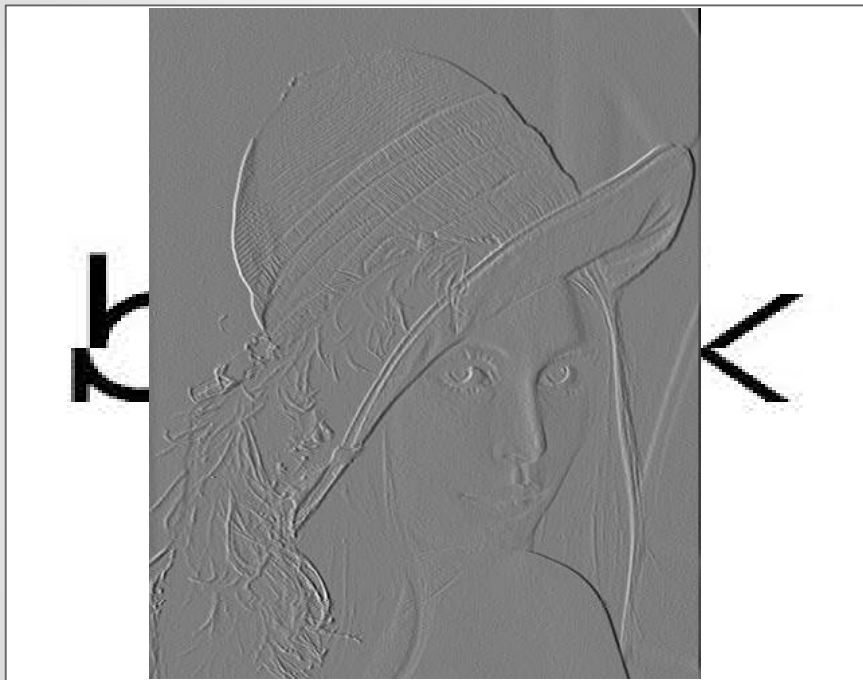
ps1-4-b-1.png

# 4c: Shifted Image



ps1-4-c-1.png

# 4d: Difference Image



ps1-4-d-1.png

The left image is intensity-scaled in the difference calculation. An image without scaling is stored in the “temp” folder. As we see, the scaled one retains more information of edges than the unscaled one.

# 5a: Noisy Green Channel



ps1-5-a-1.png

[What value of Sigma is used?]

The value of Sigma I used is 10. With this value, there are obvious noises in the image as you can see the skin is not very smooth.

This number may change depending on different images, people, and monitors.

Note: After adding (+/-) noise to green channel, the pixel values clips at 0 and 255. Although rescaling the intensity is also doable, I think for an image sensor once noises overflow its limits, it is reasonable to clip at the limits since noises intend to do this.

## 5b: Noisy Blue Channel



ps1-5-b-1.png

## 5c: Comparing Noise

Noisy blue channel image looks better to me since it is closer to the original image and losses less information. However, noisy green channel is more “rough”. This may be because the visual system of most people including me is more sensitive to the green color (eventhough the standard deviaiton of green channel is smaller than the blue channel in this image.). Thus, for the same noises, people will feel more tiny pixel intensity change in green channel than blue channel.

The results may vary depending on different monitors.