Comp Photography (Sp 2015)
HW 5

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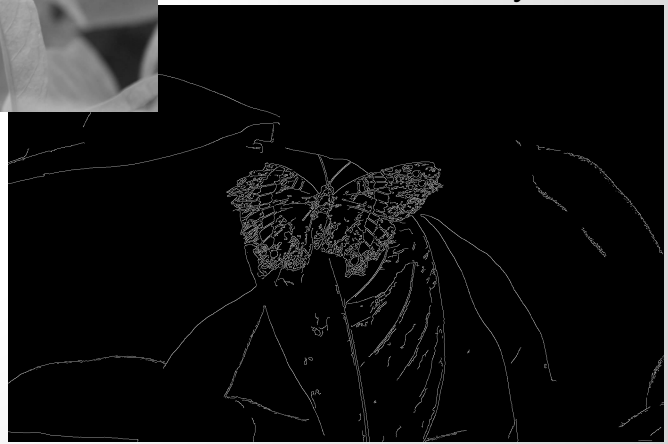
Gradient Functions

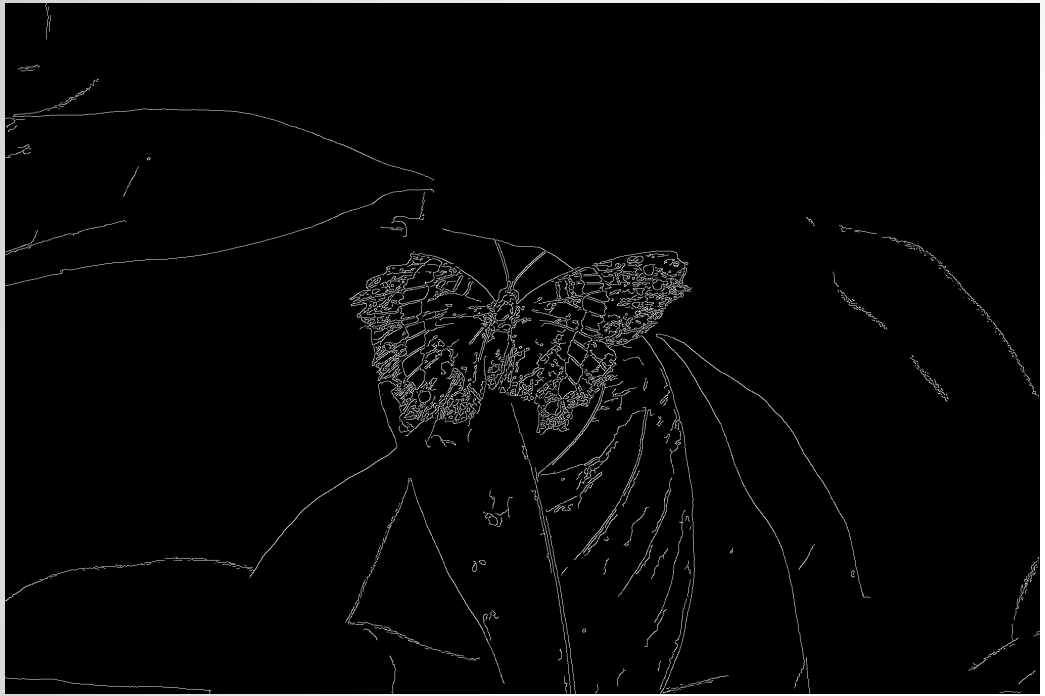
1. imageGradientX
   a. The simplest and fastest way of accomplishing this is to use the numpy “roll” function to shift all pixels to the right (with wrapping).
   b. But because we are supposed to iterate over the pixels, I converted image to int, looped through y, then x, setting the result value to:
      abs(image[y, x]-image[y,(x+1)%image.shape[1]]
   c. Convert back to uint8 (no need to worry about overflow because the biggest absolute delta will be 255)
2. imageGradientY
   a. Same as “imageGradientX” only use:
   b. abs(image[y, x]-image[(y+1)%image.shape[0],x])
3. computeGradient
   a. Convert to float
   b. The simplest and fastest way is to use “scipy.signal.convolve2d”
   c. But because other libraries are disallowed I loop through y, then x (starting at 1 ending 1 less than image.shape)
   d. set result equal to sum of numpy.multiply of slice around y,x and the kernel

Overview

Kernel 1
7x7 Gaussian
Threshold 0.05

Smoothed sigma 1
Canny 15, 45



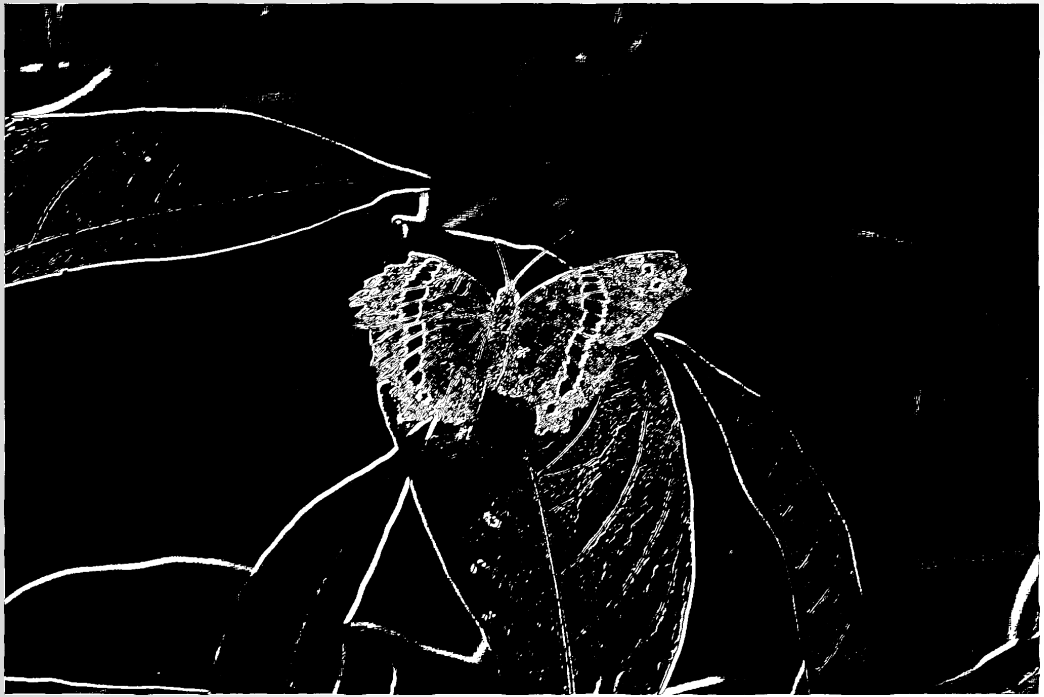


Canny

First I smoothed the image using OpenCV “GaussianBlur” (7x7 with sigma of 1)

Then I used the OpenCV function “Canny” with threshold1 = 15 and threshold2 = 45.

Notice the single pixel width.



Single Kernel

To create my own edges I tried a variety of kernels convolving a differential and gaussian kernel and then applying.

The differentials I tried were:

0 -1	-1 -1	-1 -2
-1 2	-1 3	-2 5

(with the other rows and columns 0)

The gaussians ranged from 3x3, 5x5, 7x7 with sigma 0.5, 1, 2.

I then convolved the kernel with the image. Then use absolute value of difference between pixel and median (distance from median). Normalize these values based on the maximum. Then threshold to 0 or 255. (Some nifty numpy matrix functions make all of this expedient!)
This image is pretty close (biggest difference is the edge width)