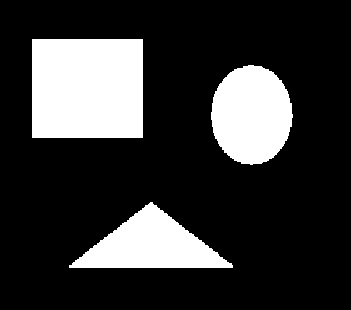
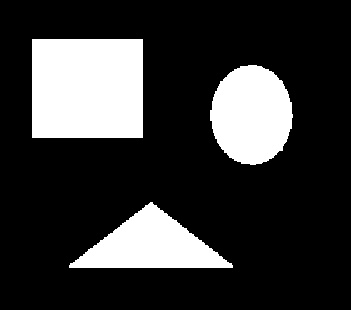
Project 2 of CSE 473/573

Shaoming Xu UB# 50247057

# Task 1

Question1

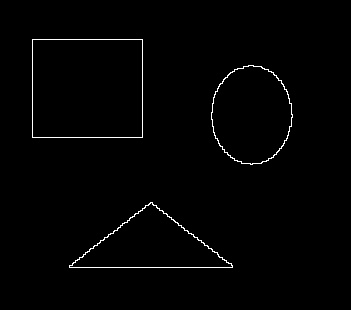
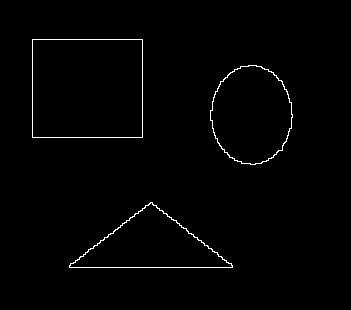
 

res\_noise1.jpg res\_noise2.jpg

Question2

They are different in my report. The res\_noise1.jpg is got from first opening then closing. The res\_noise2.jpg is got from first closing then opening. Therefore, the size of the object in res\_noise2 will be larger or equal to the res\_noise1. We can find the size of ellipses and triangles in res\_noise2 are larger than those size in res\_noise1.

Question3

res\_bound1.jpg res\_bound2.jpg

## Task1 Source code:

def dilation(binary\_image, selem):

row\_pad = math.floor(selem.shape[0] / 2)

col\_pad = math.floor(selem.shape[1] / 2)

res\_img = np.zeros(binary\_image.shape)

check\_img = np.pad(binary\_image, ((row\_pad,row\_pad),(col\_pad, col\_pad)), 'constant')

flipped\_selem = np.flip(selem)

for i in np.arange(row\_pad, check\_img.shape[0] - row\_pad):

for j in np.arange(col\_pad, check\_img.shape[1] - col\_pad):

patch = check\_img[i-row\_pad:i+row\_pad+1, j-col\_pad:j+col\_pad+1]

if np.any(patch & flipped\_selem == 1):

res\_img[i-row\_pad, j-col\_pad] = 1

return res\_img.astype(np.uint8)

def erosion(binary\_image, selem):

row\_pad = math.floor(selem.shape[0] / 2)

col\_pad = math.floor(selem.shape[1] / 2)

res\_img = np.zeros(binary\_image.shape)

check\_img = np.pad(binary\_image, ((row\_pad,row\_pad),(col\_pad, col\_pad)), 'constant')

for i in np.arange(row\_pad, check\_img.shape[0] - row\_pad):

for j in np.arange(col\_pad, check\_img.shape[1] - col\_pad):

patch = check\_img[i-row\_pad:i+row\_pad+1, j-col\_pad:j+col\_pad+1]

if np.all(patch & selem == selem):

res\_img[i-row\_pad, j-col\_pad] = 1

return res\_img.astype(np.uint8)

def opening(binary\_image, selem):

return dilation(erosion(binary\_image, selem), selem)

def closing(binary\_image, selem):

return erosion(dilation(binary\_image, selem), selem)

def boundary(binary\_image):

selem = np.ones((3,3)).astype(np.uint8)

return binary\_image - erosion(binary\_image, selem)

def denoising(method=1):

def method\_1(binary\_image, selem):

return closing(opening(binary\_image, selem), selem)

def method\_2(binary\_image, selem):

return opening(closing(binary\_image, selem), selem)

if method == 1:

return method\_1

else:

return method\_2

def threshold(gray\_img, thresh):

res\_img = np.zeros(gray\_img.shape)

for i in range(gray\_img.shape[0]):

for j in range(gray\_img.shape[1]):

if gray\_img[i,j] > thresh:

res\_img[i,j] = 1

return res\_img.astype(np.uint8)

if \_\_name\_\_ == "\_\_main\_\_":

img = cv2.imread("../task1\_img/noise.jpg")

img = cv2.imread("../task1\_img/noise.jpg")

gray = cv2.cvtColor(img,cv2.COLOR\_BGR2GRAY)

thresh = 127

struc\_elem = np.ones((3,3)).astype(np.uint8)

binary = threshold(gray, thresh)

print("\*\*\*\* task1 question a \*\*\*\*")

res\_noise1 = denoising(method=1)(binary, struc\_elem)

name = "../task1\_img" + "/res\_noise1" + ".jpg"

cv2.imwrite(name, res\_noise1\*255)

res\_noise2 = denoising(method=2)(binary, struc\_elem)

name = "../task1\_img" + "/res\_noise2" + ".jpg"

cv2.imwrite(name, res\_noise2\*255)

print("\*\*\*\* task1 question c \*\*\*\*")

res\_bound1 = boundary(res\_noise1)

name = "../task1\_img" + "/res\_bound1" + ".jpg"

cv2.imwrite(name, res\_bound1\*255)

res\_bound2 = boundary(res\_noise2)

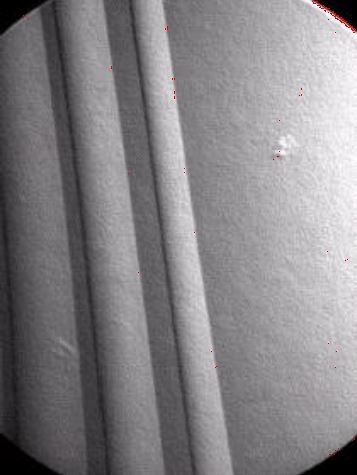
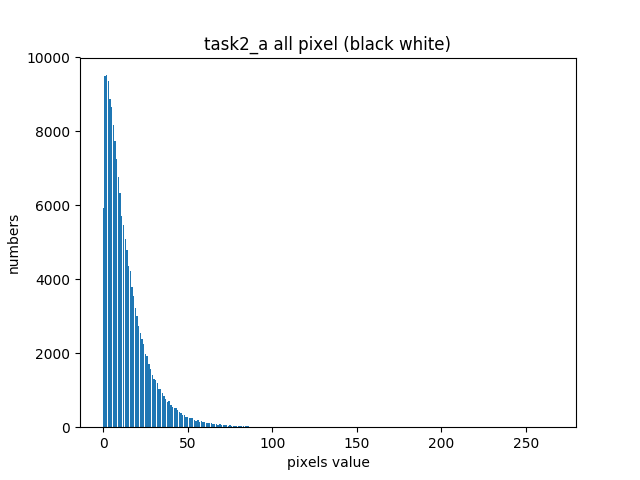
name = "../task1\_img" + "/res\_bound2" + ".jpg"

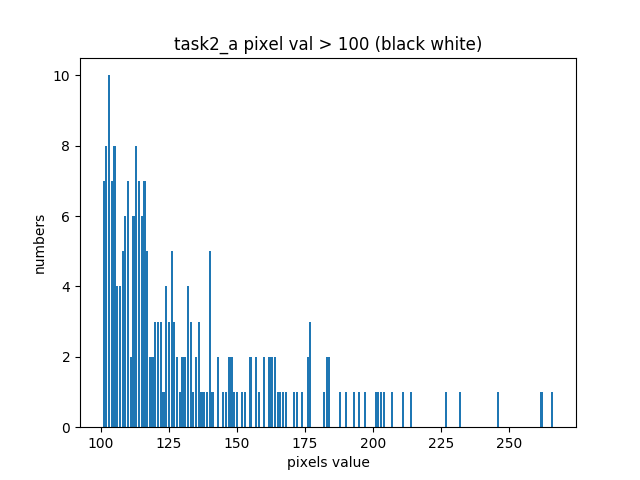
cv2.imwrite(name, res\_bound2\*255)

# Task 2a:

In this part, I have done two set experiments. One is to do point detection for both white and black point as same as that in slide. The other is only detect the white point. It turns out in our case, only do white point detection is better. Below I will discuss these two separately.

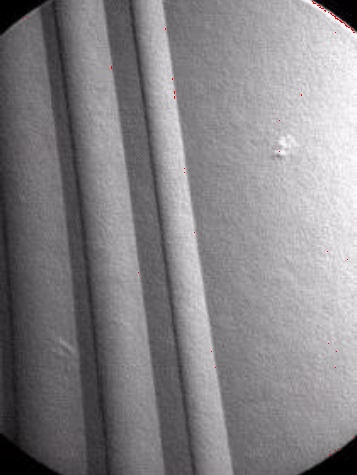
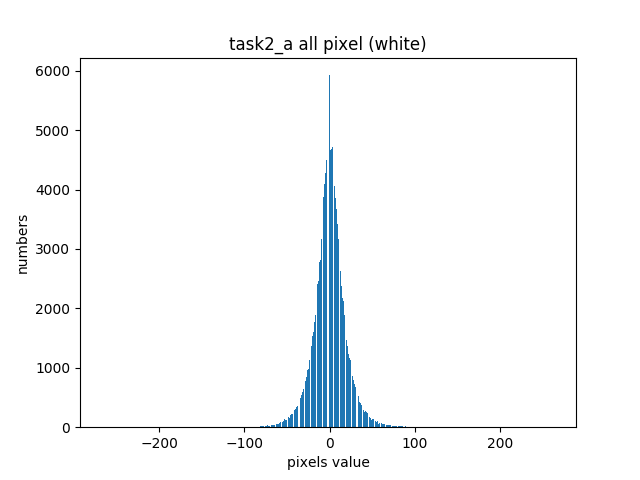
Let’s discuss white and black case First.

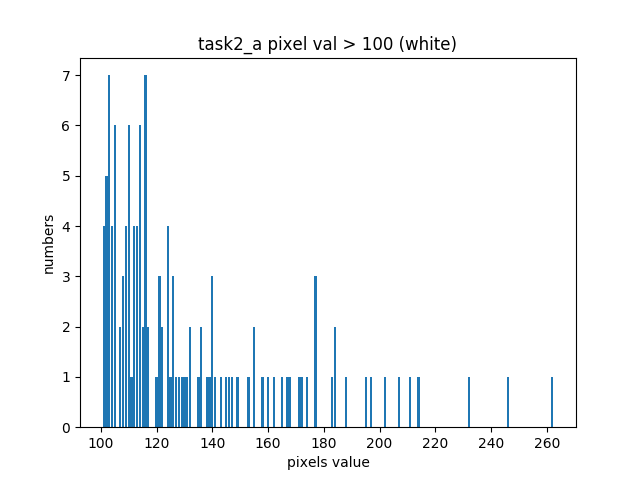
 



The left image upon show the result after white black point detection. The left two images are the histogram for the pixel value counts on absolute matrix of the original image convoluted with point detection mask. The threshold 110 is got from observation from histogram graph.

Then here we discuss the white case. The only difference of it is that we do not do absolute operation on the convoluted image. We can see the histogram has the pixel values smaller than zero.





The left image upon show the result in this case. We can see it has less noise than the first one. The threshold is 110, which is got from observation on histogram graph and experiments.

# Task 2a code:

import numpy as np

import cv2

import matplotlib.pyplot as plt

from mylibrary import count\_pixels

def get\_masked\_img(img\_gray):

"""

Purpose:

measure the weighted difference between the center point and its neighbors.

Input:

img\_gray: matrix, value type can be real.

Output:

res\_img: matrix, value type can be real.

"""

mask = np.array([[-1,-1,-1],

[-1,8,-1],

[-1,-1,-1]])

pad\_img = np.pad(img\_gray, ((1,1),(1,1)), 'constant')

res\_img = np.zeros(img\_gray.shape)

for i in range(1, pad\_img.shape[0] - 1):

for j in range(1, pad\_img.shape[1] - 1):

res\_img[i-1,j-1] = np.sum(pad\_img[i-1:i+2, j-1:j+2] \* mask)

return res\_img

def detect\_point(img\_gray, T, white\_only=False):

"""

Purpose:

Detect points in images

Input:

img\_gray: matrix with value type real

T: threshold, used to record pixels which value greater T.

white\_only: boolean, if False, detect black and white points. Otherwise, only detect white.

Output:

res\_img: matrix with value type np.uint8. use to record points.

stat: a dictionary, key is pixel value, value is number of given pixel value.

"""

masked\_img = get\_masked\_img(img\_gray)

if white\_only == False:

masked\_img = np.abs(masked\_img)

stat = count\_pixels(masked\_img, include\_border=False)

res\_img = np.zeros(img\_gray.shape).astype(np.uint8)

#ignore border

for i in range(1, masked\_img.shape[0] - 1):

for j in range(1, masked\_img.shape[1] - 1):

if masked\_img[i,j] > T:

res\_img[i,j] = 1

return res\_img, stat

def mark\_img(img, binary\_img):

res = img.copy()

for i in range(binary\_img.shape[0]):

for j in range(binary\_img.shape[1]):

if binary\_img[i,j] == 1:

res[i,j] = np.array([0,0,255]).astype(np.uint8)

return res

if \_\_name\_\_ == "\_\_main\_\_":

## detect both black and white points

img = cv2.imread("../task2a\_img/point.jpg")

gray\_img = cv2.cvtColor(img, cv2.COLOR\_BGR2GRAY)

binary, stat = detect\_point(gray\_img, 110)

res\_img = mark\_img(img, binary)

key\_val\_arr = np.asarray([[key, val] for key,val in stat.items()])

check\_all = key\_val\_arr.T

plt.bar(check\_all[0],check\_all[1],align='center') # A bar chart

plt.title('task2\_a all pixel (black white)')

plt.xlabel('pixels value')

plt.ylabel('numbers')

plt.savefig("../task2a\_img/task2a\_hist\_all\_black\_white")

plt.close()

check\_greater = np.asarray([x for x in key\_val\_arr if x[0] > 100]).T

plt.bar(check\_greater[0],check\_greater[1],align='center') # A bar chart

plt.title('task2\_a pixel val > 100 (black white)')

plt.xlabel('pixels value')

plt.ylabel('numbers')

plt.savefig("../task2a\_img/task2a\_hist\_100\_black\_white")

plt.close()

cv2.imwrite('../task2a\_img/res\_point.jpg', res\_img)

## detect only white points

gray\_img = cv2.imread("../task2a\_img/point.jpg", 0)

binary, stat = detect\_point(gray\_img, 110, white\_only=True)

res\_img = mark\_img(img, binary)

key\_val\_arr = np.asarray([[key, val] for key,val in stat.items()])

check\_all = key\_val\_arr.T

plt.bar(check\_all[0],check\_all[1],align='center') # A bar chart

plt.title('task2\_a all pixel (white)')

plt.xlabel('pixels value')

plt.ylabel('numbers')

plt.savefig("../task2a\_img/task2a\_hist\_all\_white")

plt.close()

check\_greater = np.asarray([x for x in key\_val\_arr if x[0] > 100]).T

plt.bar(check\_greater[0],check\_greater[1],align='center') # A bar chart

plt.title('task2\_a pixel val > 100 (white)')

plt.xlabel('pixels value')

plt.ylabel('numbers')

plt.savefig("../task2a\_img/task2a\_hist\_100\_white")

plt.close()

cv2.imwrite('../task2a\_img/res\_point\_white.jpg', res\_img)

# Task 2b:

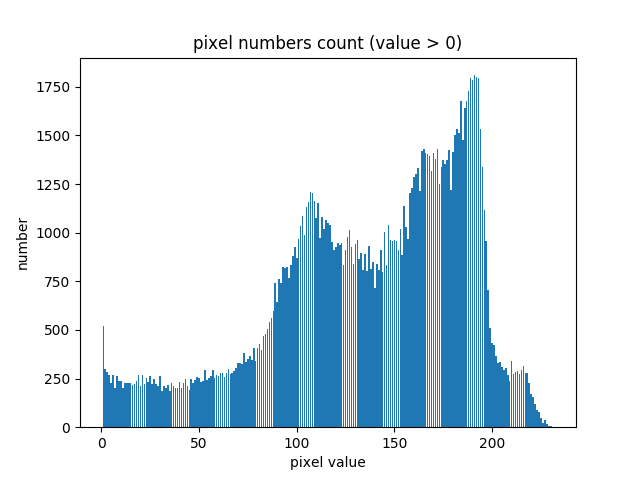
The image below shows the result.





I follow these steps to get the final results.

First, I count the pixel of gray scale original image, and use its result to draw the histogram graph.



The left is the histogram graph which ignore the pixel value of zero. By observation this graph, I do a set of experiments to get the optimal threshold as 205 to binary the image.

Second, I do the closing and then opening operation on the binary image to get the denoised image.

Third, I use DFS algorithm to recursively record all the pixel coordinations for every detected object, and save the record in the obj\_bank list.

Finally, I use the obj\_bank to draw the bounding box for each object.

Through all these steps, I got these results:

There are 10 objects

object 0 has 2314 numbers pixels

object 1 has 1273 numbers pixels

object 2 has 610 numbers pixels

object 3 has 103 numbers pixels

object 4 has 106 numbers pixels

object 5 has 89 numbers pixels

object 6 has 31 numbers pixels

object 7 has 42 numbers pixels

object 8 has 244 numbers pixels

object 9 has 140 numbers pixels

Sort by left\_up’s second value, it’s the row index. You can corelate records with objects on image easily by rows from up to down.

object 0's cordinate: left\_up=(339, 30), right,down=(362, 221)

object 1's cordinate: left\_up=(390, 44), right,down=(408, 162)

object 2's cordinate: left\_up=(255, 81), right,down=(295, 132)

object 3's cordinate: left\_up=(184, 128), right,down=(197, 137)

object 4's cordinate: left\_up=(166, 148), right,down=(177, 159)

object 5's cordinate: left\_up=(189, 152), right,down=(199, 161)

object 6's cordinate: left\_up=(263, 195), right,down=(269, 200)

object 7's cordinate: left\_up=(345, 223), right,down=(347, 236)

object 8's cordinate: left\_up=(398, 231), right,down=(419, 248)

object 9's cordinate: left\_up=(342, 270), right,down=(357, 284)

# Task 2b code:

import numpy as np

import cv2

import matplotlib.pyplot as plt

import task1

from mylibrary import count\_pixels

import sys

def thresh(gray\_img, T):

res\_img = np.zeros(gray\_img.shape).astype(np.uint8)

for i in range(gray\_img.shape[0]):

for j in range(gray\_img.shape[1]):

if gray\_img[i,j] > T:

res\_img[i,j] = 1

return res\_img

def numIslands(binary\_img):

if binary\_img is None or binary\_img.shape[0] == 0 or binary\_img.shape[1] == 0:

return 0, []

obj\_bank = []

binary = binary\_img.copy()

for i in range(binary.shape[0]):

for j in range(binary.shape[1]):

if binary[i,j] == 1:

obj = []

dfs(binary, i, j, obj)

obj\_bank.append(obj)

return obj\_bank

def dfs(grid, x, y, obj):

n = grid.shape[0]

m = grid.shape[1]

if x < 0 or x > n - 1 or y < 0 or y > m - 1 or grid[x, y] == 0:

return

obj.append([x,y])

grid[x,y] = 0

dfs(grid, x+1, y, obj)

dfs(grid, x-1, y, obj)

dfs(grid, x, y-1, obj)

dfs(grid, x, y+1, obj)

def draw\_box(img, binary, obj\_bank):

res\_img = img.copy()

binary\_color = cv2.cvtColor(binary\*255, cv2.COLOR\_GRAY2BGR)

res\_dcolor = binary\_color

for i, obj in enumerate(obj\_bank):

obj = np.asarray(obj).T

up = min(obj[0])

down = max(obj[0])

left = min(obj[1])

right = max(obj[1])

print("object {}'s cordinate: left\_up={}, right,down={}".format(i, (left, up), (right, down)))

res\_img = cv2.rectangle(res\_img,(left, up),(right,down),(0,255,0),shift=0)

res\_dcolor = cv2.rectangle(res\_dcolor,(left, up),(right,down),(0,255,0),shift=0)

return res\_img, res\_dcolor

if \_\_name\_\_ == "\_\_main\_\_":

sys.setrecursionlimit(1500)

img = cv2.imread("../task2b\_img/segment.jpg")

img\_gray = cv2.cvtColor(img, cv2.COLOR\_BGR2GRAY)

pixel\_stat = count\_pixels(img\_gray)

stat\_list = np.asarray([[key,val] for key,val in pixel\_stat.items() if key != 0]).T

# save the histogram

plt.bar(stat\_list[0],stat\_list[1],align='center') # A bar chart

plt.title("pixel numbers count (value > 0)")

plt.xlabel('pixel value')

plt.ylabel('number')

plt.savefig("../task2b\_img/task2b\_hist")

plt.close()

# observe the histogram to get T

T = 205

binary\_img = thresh(img\_gray, 205)

#denoise the binary\_img

struc\_elem = np.ones((3,3)).astype(np.uint8)

denoised = task1.denoising(method=2)(binary\_img, struc\_elem)

obj\_bank = numIslands(denoised)

print("There are {} objects".format(len(obj\_bank)))

for i in range(len(obj\_bank)):

print("object {} has {} numbers pixels".format(i, len(obj\_bank[i])))

img\_res, dcolor\_res = draw\_box(img, denoised, obj\_bank)

cv2.imwrite('../task2b\_img/res\_segment.jpg', img\_res)

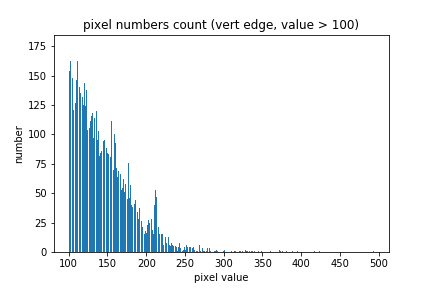
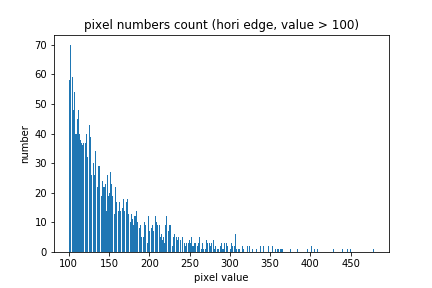
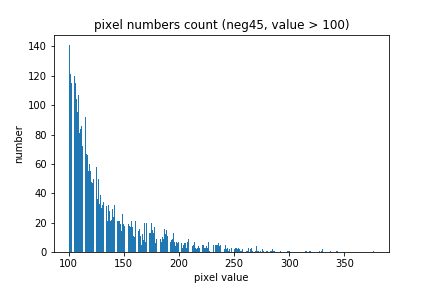
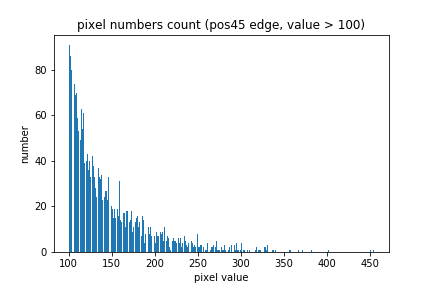
cv2.imwrite('../task2b\_img/res\_segment\_2.jpg', dcolor\_res)

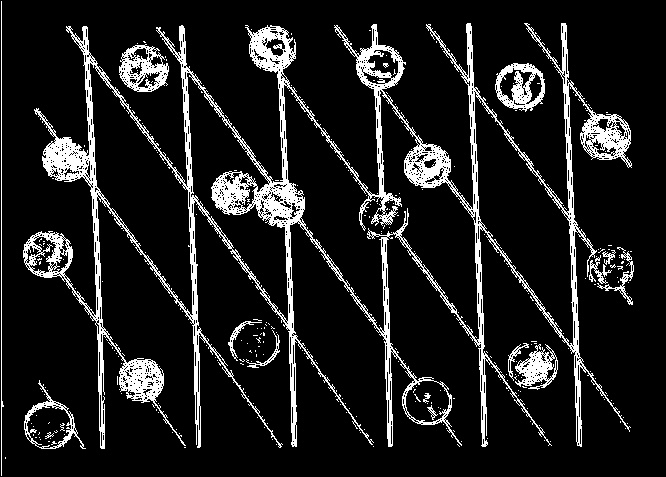
# Task 3:

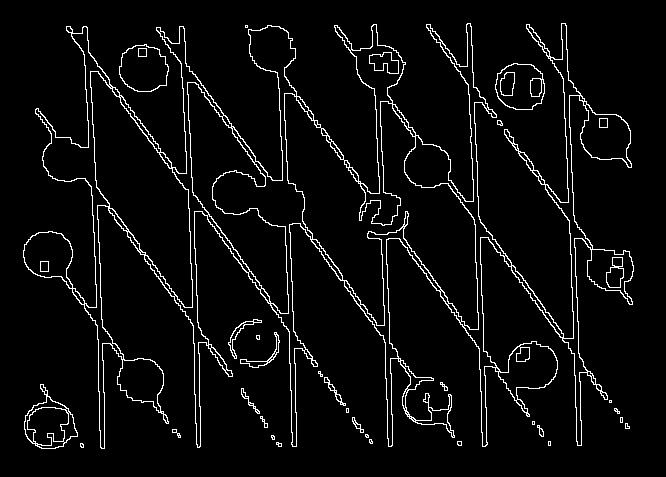
In the task3, I have implemented both edge detection algorithms and Hough transform algorithms.

Different from the opencv Canny edge detection algorithm, my edge detection is based on Morphology operation. My algorithm is followed on these steps.

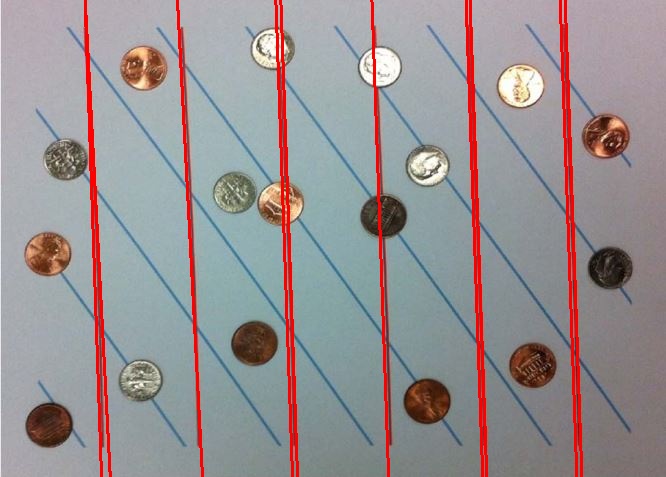
First, do the line detection on four direction, and generate four histograms.



Second, based these histograms to find the optimal threshold to combine the four edge images to a single one.

Third, do morphology operation follow these sequences: closing -> opening -> closing -> boundary to extract the boundary of image.

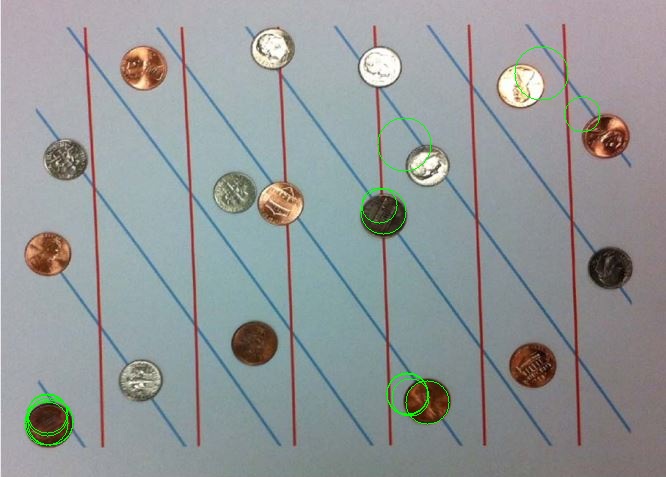
Fourth, do the Hough Transform on the boundary to find information of all Lines. To save the memory space and speed up computation, my Hough transform use a python dictionary to record the line information. The key is a tuple (rho, radian), value is the number of its key. The idea of Hough Transform algorithm is simple and elegant. It is just do voting on rho and radian space for each candidate points. Then use threshold on voting to find lines.

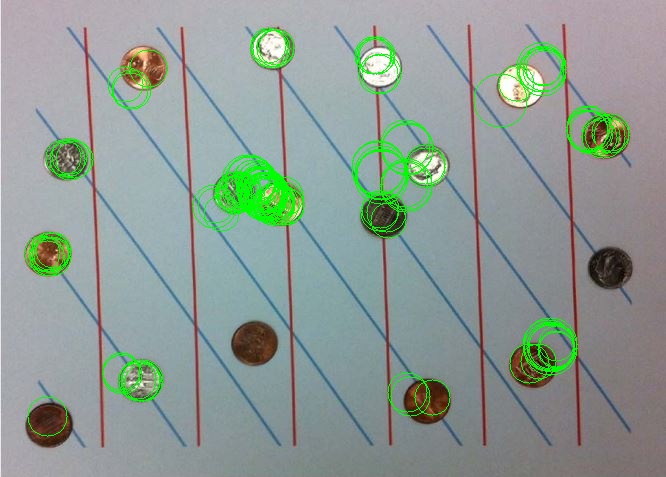
Sixth, we start to detect lines with given radians. To detect the red line, I set the radian to be 3.1 (in np.pi scale) and the gap as 0.2 to extract all line with radian in (2.9, 3.3) range. After that I set the threshold as 160 to extract lines with Hough rating over 160. Then I draw the lines on image. The left image shows I detect all red lines correctly.

To detect the blue lines. I set the radian as 2.5 and gap as 0.2 to extract all lines in with radian in (2.3, 0.2) range. After that I set the threshold as 140 to extract lines with Hough rating over 140. Then I draw the lines on image. The left image shows I detect six blue lines.

# Task 3 bonus:

I use two method to do circle detection. One is to do circle detection on my edge detection algorithm. The other is to directly do the circle detection on opencv canny edge. The canny edge method can detect more circle, but it is very slow. The idea behind Hough circle algorithm is also very simple and elegant. It is just do voting on (a, b, r) for each candidate points. Then use threshold on voting to find circle (a, b are the center of circle, r is the radius). There are two kinds of coins, I set the r range in 18 to 28 in my algorithm.

The left image shows the result of the circle detected by using the Hough circle on my own edge detection. It detects three coins correctly.

This left image shows the result of the circle detected by using on the canny edge detection. We can see it detect 15 coins, along with many incorrect circles.

At last, this method is very computation intensive. I recommend you to comment it out in my code if you time is limited.

# Task 3 code:

import numpy as np

import cv2

import math

import matplotlib.pyplot as plt

from mylibrary import count\_pixels

import task1

## the edge detection part

# sober for edge detection

VERTICAL\_SOBEL\_3BY3 = np.array([[-1,2,-1],

[-1,2,-1],

[-1,2,-1]])

HORIZONTAL\_SOBEL\_3BY3 = np.array([[-1,-1,-1],

[2,2,2],

[-1,-1,-1]])

POS45\_SOBEL\_3BY3 = np.array([[-1,-1,2],

[-1,2,-1],

[2,-1,-1]])

NEG45\_SOBEL\_3BY3 = np.array([[2,-1,-1],

[-1,2,-1],

[-1,-1,2]])

def texture\_filtering(img\_gray, kernel):

"""

Purpose:

use to filter the gray image given the kernel

Input:

img\_gray:

an two dimension ndarray matrix, dtype:usually is uint8 representint the gray image.

kernel:

a two dimension ndarray matrix

Output:

The filtered image without padding around.

"""

row\_pad = math.floor(kernel.shape[0] / 2)

col\_pad = math.floor(kernel.shape[1] / 2)

res\_img = np.zeros(img\_gray.shape)

check\_img = np.pad(img\_gray, ((row\_pad,row\_pad),(col\_pad, col\_pad)), 'constant')

flipped\_kernel = np.flip(kernel)

for i in range(row\_pad, check\_img.shape[0] - row\_pad):

for j in range(col\_pad, check\_img.shape[1] - col\_pad):

patch = check\_img[i-row\_pad:i+row\_pad+1, j-col\_pad:j+col\_pad+1]

res\_img[i-row\_pad, j-col\_pad] = np.sum(patch \* flipped\_kernel)

return res\_img

def eliminate\_zero(img, method=1):

"""

Purpose:

two ways to eliminate the negative value or the value out of 255.

Input:

img: two dimension matrix

the raw image. dtype usually is float64 with pixel < 0 or pixel > 255

method: int

default is 1 which directs to first method

the 2 will direct to the second method.

Output:

a matrix dtype range zero to one.

"""

if method == 1:

min\_ = np.min(img)

max\_ = np.max(img)

return (img - min\_) / (max\_ - min\_)

elif method == 2:

abs\_img = np.abs(img)

return abs\_img / np.max(abs\_img)

else :

print("method is 1 or 2")

def combine\_edge(hori,vert,pos45,neg45,T):

"""

Purpose:

to combine edges in four direction.

Mark result image 1 in given loc if any of these four edge images pixels absolute value

greater than T in that loc.

Output:

res\_img: binary img, value type np.uint8

"""

hori = np.abs(hori)

vert = np.abs(vert)

pos45 = np.abs(pos45)

neg45 = np.abs(neg45)

res\_img = np.zeros(hori.shape).astype(np.uint8)

for i in range(hori.shape[0]):

for j in range(hori.shape[1]):

if hori[i,j] > T or vert[i,j] > T or pos45[i,j] > T or neg45[i,j] > T :

res\_img[i,j] = 1

return res\_img

def ignore\_border(img\_gray):

"""

Purpose:

ignore the img bodre by setting value as 0.

notice: Affect original img.

Output:

the pointer point to original image.

"""

img\_gray[0] = 0

img\_gray[-1] = 0

for row in img\_gray:

row[0] = 0

row[-1] = 0

return img\_gray

### the hough algorithms part

def hough(edge\_img):

"""

Purpose:

Main function of hough transform.

Notice: Here I use the rho = x \* sin(theta) + y \* cos(theta). It is different from the hough equation.

I do this to meet the convention of opencv so that to use opencv library easily later.

Input:

edge\_img: binary image, the boundary image after edge detection.

Output:

res\_dic: a dictionary, key is a tuple (rho, radian), value is the number of its key.

"""

m, n = edge\_img.shape

res\_dic = {}

for i in range(m):

for j in range(n):

if edge\_img[i,j] == 1:

for degree in range(0,180):

radian = np.radians(degree)

rho = int(i \* np.sin(radian) + j \* np.cos(radian))

if ((rho, radian)) not in res\_dic:

res\_dic[(rho, radian)] = 0

res\_dic[(rho, radian)] += 1

return res\_dic

def pick\_by\_count(lines, T):

"""

Purpose:

filter the lines which number of (rho, radian) greater than threshold T.

Input:

lines: a matrix, col\_1: rho, col\_2: radian, col\_2: number of (rho, radian) tuple. It can be got by res\_dic.

Output:

matrix, numpy array, each row represent a selected line.

"""

res = []

for line in lines:

if line[2] > T:

res.append(line)

return np.asarray(res)

def pick\_by\_theta(lines, theta, gap):

"""

Purpose:

filter the lines by theta,

Input:

lines: a matrix, col\_1: rho, col\_2: radian, col\_2: number of (rho, radian) tuple.

theta: real, is a radian, will compared radians from col\_2

gap: the maximum difference between two theta.

Output:

matrix, numpy array, each row represent a selected line.

"""

res = []

for line in lines:

if abs(line[1] - theta) < gap:

res.append(line)

return np.asarray(res)

def draw\_lines(img, lines, loc="../task3\_img/", name = "hough\_res.jpg"):

"""

Purpose:

draw lines in image

not affect original imgage.

"""

img = img.copy()

for line in lines:

rho = line[0]

theta = line[1]

a = np.cos(theta)

b = np.sin(theta)

x0 = a\*rho

y0 = b\*rho

x1 = int(x0 + 1000\*(-b))

y1 = int(y0 + 1000\*(a))

x2 = int(x0 - 1000\*(-b))

y2 = int(y0 - 1000\*(a))

cv2.line(img,(x1,y1),(x2,y2),(0,0,255),2)

cv2.imwrite(loc+name, img)

def circle\_hough(edge\_img):

res\_dic = {}

for x in range(edge\_img.shape[0]):

for y in range(edge\_img.shape[1]):

if edge\_img[x,y] == 1:

for r in range(18,28):

for t in range(0,360):

a = x - r \* np.cos(t \* np.pi / 180)

b = y - r \* np.sin(t \* np.pi / 180)

if (a,b,r) not in res\_dic:

res\_dic[(a,b,r)] = 0

res\_dic[(a,b,r)] += 1

return res\_dic

def draw\_circles(img, circles, loc="../task3\_img/", name="coin.jpg" ):

img = img.copy()

for cir in circles:

cv2.circle(img,(cir[1], cir[0]), cir[2], (0,255,0), 1)

cv2.imwrite(loc+name, img)

if \_\_name\_\_ == "\_\_main\_\_":

img = cv2.imread('../task3\_img/hough.jpg')

img\_gray = cv2.cvtColor(img, cv2.COLOR\_BGR2GRAY)

print("Start to do edge detection")

## get edges from four directions

vertical = texture\_filtering(img\_gray, VERTICAL\_SOBEL\_3BY3)

horizontal = texture\_filtering(img\_gray, HORIZONTAL\_SOBEL\_3BY3)

pos45 = texture\_filtering(img\_gray, POS45\_SOBEL\_3BY3)

neg45 = texture\_filtering(img\_gray, NEG45\_SOBEL\_3BY3)

## Find threshold through observation

# vertical

pixel\_stat = count\_pixels(np.abs(vertical))

stat\_list = np.asarray([[key,val] for key,val in pixel\_stat.items() if key >100]).T

plt.bar(stat\_list[0],stat\_list[1],align='center') # A bar chart

plt.title("pixel numbers count (vert edge, value > 100)")

plt.xlabel('pixel value')

plt.ylabel('number')

plt.savefig("../task3\_img/vert\_hist")

plt.close()

#horizontal

pixel\_stat = count\_pixels(np.abs(horizontal))

stat\_list = np.asarray([[key,val] for key,val in pixel\_stat.items() if key >100]).T

plt.bar(stat\_list[0],stat\_list[1],align='center') # A bar chart

plt.title("pixel numbers count (hori edge, value > 100)")

plt.xlabel('pixel value')

plt.ylabel('number')

plt.savefig("../task3\_img/hori\_hist")

plt.close()

#pos45

pixel\_stat = count\_pixels(np.abs(pos45))

stat\_list = np.asarray([[key,val] for key,val in pixel\_stat.items() if key >100]).T

plt.bar(stat\_list[0],stat\_list[1],align='center') # A bar chart

plt.title("pixel numbers count (pos45 edge, value > 100)")

plt.xlabel('pixel value')

plt.ylabel('number')

plt.savefig("../task3\_img/pos45\_hist")

plt.close()

#neg45

pixel\_stat = count\_pixels(np.abs(neg45))

stat\_list = np.asarray([[key,val] for key,val in pixel\_stat.items() if key >100]).T

plt.bar(stat\_list[0],stat\_list[1],align='center') # A bar chart

plt.title("pixel numbers count (neg45, value > 100)")

plt.xlabel('pixel value')

plt.ylabel('number')

plt.savefig("../task3\_img/neg45\_hist")

plt.close()

## Combined edges using threshold

T = 50

combined = combine\_edge(horizontal,vertical,pos45,neg45,T)

combined = ignore\_border(combined)

cv2.imwrite("../task3\_img/combined.jpg", (combined\*255).astype(np.uint8))

## preprocess binary image

print("Start to preprocess binary image")

#denoise

struc\_elem = np.ones((3,3)).astype(np.uint8)

denoised = task1.denoising(method=2)(combined, struc\_elem)

cv2.imwrite("../task3\_img/denoised.jpg", (denoised\*255).astype(np.uint8))

#closing

struc\_elem = np.ones((7,7)).astype(np.uint8)

closed = task1.closing(denoised, struc\_elem)

cv2.imwrite("../task3\_img/closed.jpg", (closed\*255).astype(np.uint8))

#extract boundary

boundary = task1.boundary(closed)

cv2.imwrite("../task3\_img/boundary.jpg", (boundary\*255).astype(np.uint8))

## Start using hough algorithms

print("Start using hough algorithms")

hres = hough(boundary)

lines = np.asarray([[key[0], key[1], val] for key, val in hres.items()])

#filter by theta and count

#red lines

T = 160

theta\_gap = 0.2

theta\_red = 3.1

lines\_red = pick\_by\_theta(lines, theta\_red, theta\_gap)

res\_red = pick\_by\_count(lines\_red, T)

draw\_lines(img, res\_red, loc="../task3\_img/", name = "red\_line.jpg")

#blue lines

T = 140

theta\_gap = 0.2

theta\_blue = 2.5

lines\_blue = pick\_by\_theta(lines, theta\_blue, theta\_gap)

res\_blue = pick\_by\_count(lines\_blue, T)

draw\_lines(img, res\_blue, loc="../task3\_img/", name = "blue\_lines.jpg")

#Bonus Circle hough

print("start hough circle")

chres = circle\_hough(boundary)

circles = [item for item in chres.items() if chres[item[0]] >= 4]

circles = np.asarray([[x[0][0], x[0][1], x[0][2], x[1]] for x in circles]).astype(int)

draw\_circles(img, circles)

#Bonus using Canny edge detection

print("hough circle on Canny edge")

img = cv2.imread('../task3\_img/hough.jpg')

gray = cv2.cvtColor(img,cv2.COLOR\_BGR2GRAY)

edges = cv2.Canny(gray,50,150,apertureSize = 3)

edges = (edges/255).astype(np.uint)

chres = circle\_hough(edges)

circles = [item for item in chres.items() if chres[item[0]] >= 4]

circles = np.asarray([[x[0][0], x[0][1], x[0][2], x[1]] for x in circles]).astype(int)

draw\_circles(img, circles, name="coin\_edge.jpg")