## Computer Vision — Homework 2

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### 1 A binary image (threshold at 128)

```
# binarize lena.bmp at 128 to get a binary image
 9
    binarize = np.zeros(image.shape, int)
10
    for i in range(image_rows):
11
         for j in range(image_cols):
12
             if image[i][j] < 128:</pre>
13
                 binarize[i][j] = 0
14
             else:
15
                 binarize[i][j] = 255
16
    cv2.imwrite('binarize.jpg', binarize)
17
```

Actually, the picture is already grayscale, so we don't need to transfer the color into grayscale again. Then, identify the value. If it is less than 128, change it to black (pixel = 0), else, change it to white (pixel =255).



Figure 1: binarize at threshold 128

#### 2 A histogram

```
# draw a histogram
19
    histogram = np.zeros(256, int)
20
    index = np.arange(256)
21
22
    for i in range(image_rows):
23
        for j in range(image_cols):
             histogram[image[i][j]] += 1
24
    plt.bar(index, histogram)
25
    plt.ylabel("Counts")
26
    plt.xlabel('Gray Level')
27
    plt.title('HISTOGRAM OF LENA.BMP')
28
29
    plt.show()
```

Create an array and make it size 256. This array represents the number of each gray value from 0 to 255 on lena.bmp. Check each pixel's gray value on lena.bmp and make that gray value's count +1.

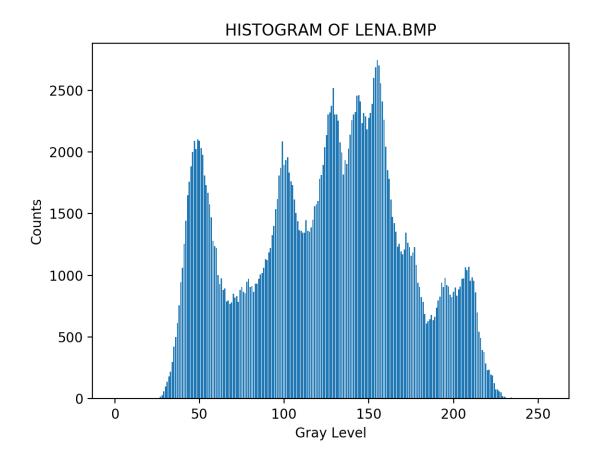


Figure 2: Histogram

# 3 Connected components (regions with + at centroid, bounding box)

```
for i in range(image3_rows):
        row_counts = 0
        startCOL =
43
        endCOL = 0
        for j in range(image3_cols):
45
46
             if image3[i][j] > 128: # pixel 0 黑色, pixel != 0 白色
47
                 if in_run == 0:
48
                     in run = 1
                     startCOL = j
49
50
                     row counts += 1
51
                        = image3_cols - 1:
52
                     in_run = 0
53
                     endCOL = j
                     run_data.append([i, startCOL, endCOL, 0, endCOL - startCOL + 1])
54
55
             else:
56
                    in_run != 0:
57
                     in_run = 0
58
                     endCOL = j - 1
                     run_data.append([i, startCOL, endCOL, 0, endCOL - startCOL + 1])
59
60
61
        if row_counts == 0:
             counts_inRow.append([0, 0])
62
        else:
63
             length = len(run_data)
64
65
             counts_inRow.append([length - row_counts + 1, length])
```

The first loop is responsible for recording two lists, counts\_inRow and run\_data.

counts\_inRow: [0]: start of the run, [1]: end of the run

*run\_data*: [0]: this run on which row, [1]: this run starting from which column, [2]: ending from which column, [3]: label, [4]: the number of pixels in this run

(a) Top-down pass

(b) Bottom-up pass

Figure 3: Run Length Implementation

Implement the third method on PowerPoint – Run Length Implementation. Maintain *label* list at the same time when tracing top-down pass and bottom-up pass steps.

label: Its index represents label number, and content represents the orders of the runs, which are in this label. For example: label=[[nothing], [0, 2, 3], [1]]. It means run #0, run #2, run #3 have the same label, label#1, and run #1 has label#2.

```
range(1, Len):
= 0
                                                                                   range(len(label[i])):
                                                                                                           run_data[label[i][j]][4]
                                                                                right_pos.append(run_data[label[i][j]][2])
left_pos.append(run_data[label[i][j]][1])
                                                                                                right_pos[i] < run_data[label[i][j]][2]:
    right_pos[i] = run_data[label[i][j]][2]
left_pos[i] > run_data[label[i][j]][1]:
    left_pos[i] = run_data[label[i][j]][1]
                         len_pixel.append(pixel)
count = 0
final_image = cv2.imread('binarize.jpg')
                       __mage = cv2.imread('
i in range(Len):
label[i].sort()
Length = len(label[i])
x_cen = 0
                                                 (len_pixel[i] >= 500):
                                                                          left_pos[i]
                                                   x = left_pos[i]
x_end = right_pos[i]
y = run_data[labet[i][0]][0]
y_end = run_data[labet[i][Length - 1]][0]
cv2.rectangle(final_image, (x, y), (x_end, y_end), (255, 255, 0), 1)
for j in range(Length):
    run_order = labet[i][j]
    x_cen_plus = ((run_data[run_order][1] + run_data[run_order][2])/ 2) * run_data[run_order][4]
    y_cen_plus = run_data[run_order][0] * run_data[run_order][4]
    x cen += x cen_plus
                                                                                                          n += y_cen_plus
len_pixel[i]
len_pixel[i]
                                                                                           = int(x_cen)
                                                                              n = int(y_cen)
line(final_image,
                                                                                                                                                                                             (x_cen - 6, y_cen), (x_cen + 6, y_cen), (0, 0, 255), thicknote (x_cen, y_cen - 6), (x_cen, y_cen + 6), (0, 0, 255), thicknote (x_cen, y_cen + 6), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255), (0, 0, 255),
                                                      cv2.line(final_image,
```

Threshold at 500, figure out the label whose area is larger than 500, use a bounding box to frame the smallest area of it, and mark the centroid of this area.



Figure 4: Result