## Computer Vision Homework #10

Student-ID: r10944020; Name: 林顥倫; Department: GINM

## [Results]

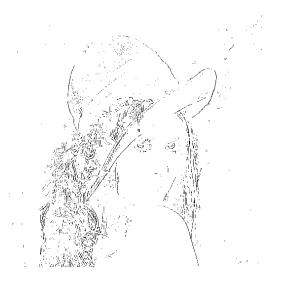
(a) Laplace Mask1 (0, 1, 0, 1, -4, 1, 0, 1, 0): 15 (b) Laplace Mask2 (1, 1, 1, 1, -8, 1, 1, 1, 1)





(c) Minimum variance Laplacian: 20

(d) Laplace of Gaussian: 3000





(e) Difference of Gaussian: 1



[Code Fragment & Explanation]

Part-1 Function for edge detection via 3\*3 kernel

針對(a)(b)(c)這三個小題,會把不同的3\*3 kernel傳入這個function進行運算。

```
30 def laplace(img, kernel, threshold):
       imageW, imageH = img.shape
32
       res = np.zeros((imageW-2, imageH-2), dtype='int32')
33
       for x in range(imageW-2):
34
           for y in range(imageH-2):
               val = convolution(img[x:x+3, y:y+3], kernel)
35
36
               if val >= threshold:
                    res[x][y] = 1
37
38
               elif val <= -threshold:</pre>
39
                    res[x][y] = -1
41
                    res[x][y] = 0
42
       return zero_crossing_detector(res, 3)
```

Part-2 Function for edge detection via 10\*10 kernel

針對(d)(e)這三個小題,會把不同的11 \* 11 kernel傳入這個function進行運算。

```
44 def edge_Gaussian(img, kernel, threshold):
45
       imageW, imageH = img.shape
       res = np.zeros((imageW-10, imageH-10), dtype='int32')
46
47
       for x in range(imageW-10):
48
           for y in range(imageH-10):
               val = convolution(img[x:x+11, y:y+11], kernel)
49
50
               if val >= threshold:
51
                    res[x][y] = 1
               elif val <= -threshold:</pre>
52
53
                    res[x][y] = -1
54
               else:
55
                    res[x][y] = 0
56
       return zero_crossing_detector(res, 3)
```

Part-3 Zero-Crossing Edge Detection 此部分依照講義公式與助教講解實作

```
13 def validPixel(x, bound):
14
       return (x >= 0 \text{ and } x <= \text{bound})
15
16 def zero_crossing_detector(img, kernel_size):
       imageW, imageH = img.shape
17
18
       res = np.full(img.shape, 255, dtype='int32')
19
       for x in range(imageW):
            for y in range(imageH):
21
                if img[x][y] == 1:
22
                    for ex in range(-kernel_size//2+1, kernel_size//2+1):
23
                         for ey in range(-kernel_size//2+1, kernel_size//2+1):
24
                             dest_x, dest_y = x + ex, y + ey
25
                             if validPixel(dest_x, imageW-1) and validPixel(dest_y, imageH-1) \setminus
26
                                 and img[dest_x][dest_y] == -1:
27
                                 res[x][y] = 0
       return res
```

Part-4 Padding 圖片傳入之前,會先做不同的Padding (for 3\*3 and 11\*11)

```
58 def padding_img(img, size):
59    return np.pad(img, (size, size), 'edge')
```

Part-5 主函式呼叫(見下頁)

```
61 if __name__ == '__main__':
         img = cv2.imread('lena.bmp', cv2.IMREAD_GRAYSCALE)
 62
 63
         img_pad1 = padding_img(img, 1)
 64
         img_pad2 = padding_img(img, 5)
 65
 66
         # (a) Laplace Mask1 (0, 1, 0, 1, -4, 1, 0, 1, 0): 15
         kernel_laplace_1 = np.array([[0, 1, 0], [1, -4, 1], [0, 1, 0]])
 67
 68
         res_a = laplace(img_pad1, kernel_laplace_1, 15)
         cv2.imwrite('res_a.png', res_a)
# (b) Laplace Mask2 (1, 1, 1, 1, -8, 1, 1, 1, 1)
 69
 70
         kernel_laplace_2 = np.array([[1, 1, 1], [1, -8, 1], [1, 1, 1]]) / 3
 71
 72
         res_b = laplace(img_pad1, kernel_laplace_2, 15)
 73
         cv2.imwrite('res_b.png', res_b)
         # (c) Minimum variance Laplacian: 20
 74
         kernel_laplace_minvar = np.array([[2, -1, 2], [-1, -4, -1], [2, -1, 2]]) / 3
 75
         res_c = laplace(img_pad1, kernel_laplace_minvar, 20)
 76
 77
         cv2.imwrite('res_c.png', res_c)
 78
         # (d) Laplace of Gaussian: 3000
 79
         kernel_LOG = np.array([
 80
              [0, 0, 0, -1, -1, -2, -1, -1, 0, 0, 0],
              [0, 0, -2, -4, -8, -9, -8, -4, -2, 0, 0],
[0, -2, -7, -15, -22, -23, -22, -15, -7, -2, 0],
 81
 82
              [-1, -4, -15, -24, -14, -1, -14, -24, -15, -4, -1],
 83
 84
              [-1, -8, -22, -14, 52, 103, 52, -14, -22, -8, -1],
              [-2, -9, -23, -1, 103, 178, 103, -1, -23, -9, -2], [-1, -8, -22, -14, 52, 103, 52, -14, -22, -8, -1],
 85
 86
              [-1, -4, -15, -24, -14, -1, -14, -24, -15, -4, -1],
 87
              [0, -2, -7, -15, -22, -23, -22, -15, -7, -2, 0], [0, 0, -2, -4, -8, -9, -8, -4, -2, 0, 0], [0, 0, 0, -1, -1, -2, -1, -1, 0, 0, 0]
 88
 89
 90
 91
 92
         res_d = edge_Gaussian(img_pad2, kernel_LOG, 3000)
 93
         cv2.imwrite('res_d.png', res_d)
 94
         # (e) Difference of Gaussian: 1
 95
         kernel_DOG = np.array([
              [-1, -3, -4, -6, -7, -8, -7, -6, -4, -3, -1],
[-3, -5, -8, -11, -13, -13, -13, -11, -8, -5, -3],
 96
 97
              [-4, -8, -12, -16, -17, -17, -17, -16, -12, -8, -4],
 98
 99
              [-6, -11, -16, -16, 0, 15, 0, -16, -16, -11, -6],
100
              [-7, -13, -17, 0, 85, 160, 85, 0, -17, -13, -7],
101
              [-8, -13, -17, 15, 160, 283, 160, 15, -17, -13, -8],
              [-7, -13, -17, 0, 85, 160, 85, 0, -17, -13, -7],
102
103
              [-6, -11, -16, -16, 0, 15, 0, -16, -16, -11, -6],
              [-4, -8, -12, -16, -17, -17, -17, -16, -12, -8, -4],

[-3, -5, -8, -11, -13, -13, -13, -11, -8, -5, -3],

[-1, -3, -4, -6, -7, -8, -7, -6, -4, -3, -1]
104
105
106
107
         ])
108
         res_e = edge_Gaussian(img_pad2, kernel_DOG, 1)
109
         cv2.imwrite('res_e.png', res_e)
```