

Problem 9

Image segmentation

问题描述

- (a). Develop a program to implement the Roberts, Prewitt, Sobel, the Marr-Hildreth and the Canny edge detectors. Use the image ‘building.tif’ to test your detectors. (For technique details of Marr-Hildreth and Canny, please refer to pp.736-747 (3rd edition, Gonzalez DIP) or MH-Canny.pdf at the same address of the slides.)
- (b). Develop a program to implement the Otsu’s method of thresholding segmentation, and compare the results with the global thresholding method using test image ‘polymersomes.tif’. (For technique details, please refer to pp.763-770 (3rd edition, Gonzalez DIP), or Otsu.pdf at the same ftp address of slides.)

算法思想

本题要求我们以 Roberts, Prewitt, Sobel, Marr - Hildreth, Canny 算子来对图像进行边缘检测。
各个算法的定义如下：

- Roberts 算子：

Roberts 算子的近似偏导定义如下：

$$g_x = z_9 - z_5$$
$$g_y = z_8 - z_6$$

- Prewitt 算子：

Prewitt 算子的近似偏导定义如下：

$$g_x = (z_7 + z_8 + z_9) - (z_1 + z_2 + z_3)$$
$$g_y = (z_3 + z_6 + z_9) - (z_1 + z_4 + z_7)$$

- Sobel 算子：

Sobel 算子的近似偏导定义如下：

$$g_x = (z_7 + 2z_8 + z_9) - (z_1 + 2z_2 + z_3)$$
$$g_y = (z_3 + 2z_6 + z_9) - (z_1 + 2z_4 + z_7)$$

在本题中，我采用了直接以偏导 g_x, g_y 相加的方式来得到近似的边缘图片。

- Marr - Hildreth 算子：

对于 MH 算子，需要先进行高斯变换和拉普拉斯变换，LoG 的定义如下：

$$LoG(x, y) = \left(\frac{x^2 + y^2 - 2\sigma^2}{\sigma^4} \right) \exp\left(-\frac{x^2 + y^2}{2\sigma^2}\right)$$

在对原始图像进行 LOG 变换后，需要先对得到的变换矩阵进行归一化。

然后对原始图片与变换后的矩阵进行卷积，求其差值，得到差值图像。

接着求出差值图像通过 0 的点，在本题中我的思路是将该点与四周四个方向的点分别求和，假如该点为 0，则必然存在一个方向上的和大于 0，且还同时有另一个方向上的和小于 0。

找出点后我进行标记，把为 0 的点当做边界。

- Canny 算子：

Canny 算子跟之前介绍的 MH 算子一样，也属于先平滑后求导数的方法。

Canny 算法的步骤如下：

- 去噪：对原始图像与高斯掩模做卷积，得到模糊图像。这样做可以达到去噪的效果。
- 用一阶偏导的有限差分来计算梯度的幅值和方向。
- 对梯度幅值进行非极大值抑制，因为仅仅得到全局的梯度还是没办法确定边缘，故需要保留局部梯度最大的点，并且抑制非极大值。
- 用双阈值算法检测和连接边缘，双阈值法的定义为：对非极大值抑制图象作用两个阈值 τ_1 和 τ_2 ，且 $2\tau_1 \approx \tau_2$ ，从而可以得到两个阈值边缘图象 $N1[i, j]$ 和 $N2[i, j]$ 。由于 $N2[i, j]$ 使用高阈值得到，因而含有很少的假边缘，但有间断(不闭合)。双阈值法要在 $N2[i, j]$ 中把边缘连接成轮廓，当到达轮廓的端点时，该算法就在 $N1[i, j]$ 的 8 邻点位置寻找可以连接到轮廓上的边缘，这样，算法不断地在 $N1[i, j]$ 中收集边缘，直到将 $N2[i, j]$ 连接起来为止。(在连接边缘的时候，用数组模拟队列的实现。以进行 8-连通域搜索)

- 全局阈值法

全局阈值法的思路如下：

- 为全局阈值选择一个初始估计值 T (图像的平均灰度)。
- 用 T 分割图像。产生两组像素： $G1$ 有灰度值大于 T 的像素组成， $G2$ 有小于等于 T 像素组成。
- 计算 $G1$ 和 $G2$ 像素的平均灰度值 $m1$ 和 $m2$ ；
- 计算一个新的阈值： $T = (m1 + m2) / 2$ ；
- 重复步骤 2 和 4，直到连续迭代中的 T 值间的差为零。

- Ostu 阈值法

在本次实验中我使用了 Ostu 大律法，算法思路如下：

一幅有 $depth$ 个灰度级，根据每个灰度级 t ，可以将一幅图分为前景和背景。前景指所有灰度级低于等于 t 的像素点，背景指大于 t 的像素点。

$w0$ 指前景像素个数； $w1$ 指背景像素个数； $u0$ 指前景加权平均，即

```
1 temp = 0;
2 for i = 1 : t
3   temp = temp + i * hist(i)
4 end
5 u0 = temp / w0;
6 u = u0 * w0 + u1 * w1
```

其中 $w0 = hist(0) + hist(1) + \dots + hist(t)$

$hist$ 指各个灰度级上的像素个数； $u1$ 对应是背景加权平均； u 对应整幅图的加权平均

大津法的结果指使得 g 最大的 t 值：

$$g = w0(u0 - u)^2 + w1(u1 - u)^2 = w0w1(u1 - u0)^2$$

再以 g 为阈值进行边缘检测。

源码分析

- 实验过程如下：

- 读取原始图片并将其转为灰度图像。

- b. 对原始图片分别使用 Roberts, Prewitt, Sobel 算子进行边缘检测，得到边缘图像，显示各种方法处理得到的边缘图像。
- c. 对原始图片使用 MH 算子进行边缘检测，实验过程在算法思想里有涉及，得到边缘图像并展示。
- d. 对原始图片使用 Canny 算子进行边缘检测，流程同算法思想，得到边缘图像并展示。
- e. 分别使用全局阈值和 Ostu 阈值进行边缘检测，得到边缘图像并展示。

```

1 def main():
2     originalImage = np.array(Image.open('./resource/building.tif'))
3
4     plt.subplot(2, 2, 1)
5     plt.imshow(originalImage, cmap = plt.get_cmap('gray'))
6     plt.title('Original')
7
8     robertsGxImage, robertsGyImage, robertsImage = segmentAlgorithm.Roberts(originalImag
e)
9     plt.subplot(2, 2, 2)
10    plt.imshow(robertsGxImage, cmap = plt.get_cmap('gray'))
11    plt.title('Roberts gx')
12
13    plt.subplot(2, 2, 3)
14    plt.imshow(robertsGyImage, cmap = plt.get_cmap('gray'))
15    plt.title('Roberts gy')
16
17    plt.subplot(2, 2, 4)
18    plt.imshow(robertsImage, cmap = plt.get_cmap('gray'))
19    plt.title('Roberts')
20
21    plt.show()
22
23    plt.subplot(2, 2, 1)
24    plt.imshow(originalImage, cmap = plt.get_cmap('gray'))
25    plt.title('Original')
26
27    prewittGxImage, prewittGyImage, prewittImage = segmentAlgorithm.Prewitt(originalImag
e)
28    plt.subplot(2, 2, 2)
29    plt.imshow(prewittGxImage, cmap = plt.get_cmap('gray'))
30    plt.title('Prewitt gx')
31
32    plt.subplot(2, 2, 3)
33    plt.imshow(prewittGyImage, cmap = plt.get_cmap('gray'))
34    plt.title('Prewitt gy')
35
36    plt.subplot(2, 2, 4)
37    plt.imshow(prewittImage, cmap = plt.get_cmap('gray'))
38    plt.title('Prewitt')
39
40    plt.show()
41
42    plt.subplot(2, 2, 1)
43    plt.imshow(originalImage, cmap = plt.get_cmap('gray'))
44    plt.title('Original')
45
46    sobelGxImage, sobelGyImage, sobelImage = segmentAlgorithm.Sobel(originalImage)
47    plt.subplot(2, 2, 2)
48    plt.imshow(sobelGxImage, cmap = plt.get_cmap('gray'))
49    plt.title('Sobel gx')
50
51    plt.subplot(2, 2, 3)

```

```

52 plt.imshow(sobelGyImage, cmap = plt.get_cmap('gray'))
53 plt.title('Sobel gy')
54
55 plt.subplot(2, 2, 4)
56 plt.imshow(sobelImage, cmap = plt.get_cmap('gray'))
57 plt.title('Sobel')
58
59 plt.show()
60
61 plt.subplot(2, 2, 1)
62 plt.imshow(originalImage, cmap = plt.get_cmap('gray'))
63 plt.title('Original')
64
65 plt.subplot(2, 2, 2)
66 LoGImage, marrHildrethImage1, marrHildrethImage2 = marrHildrethAlgorithm.MarrHildret
h(originalImage)
67 plt.imshow(LoGImage, cmap = plt.get_cmap('gray'))
68 plt.title('LoG Without Threshold')
69
70 plt.subplot(2, 2, 3)
71 plt.imshow(marrHildrethImage1, cmap = plt.get_cmap('gray'))
72 plt.title('Marr-Hildreth Threshold = 0')
73
74 plt.subplot(2, 2, 4)
75 plt.imshow(marrHildrethImage2, cmap = plt.get_cmap('gray'))
76 plt.title('Marr-Hildreth Threshold = 15% Max(LoGImage)')
77
78 plt.show()
79
80 plt.subplot(2, 2, 1)
81 plt.imshow(originalImage, cmap = plt.get_cmap('gray'))
82 plt.title('Original')
83
84 plt.subplot(2, 2, 2)
85 smoothImage, cannyImage = cannyAlgorithm.canny(Image.open('./resource/building.tif'))
86 .convert('L')
87 plt.imshow(smoothImage, cmap = plt.get_cmap('gray'))
88 plt.title('Smooth')
89
90 plt.subplot(2, 2, 3)
91 plt.imshow(marrHildrethImage2, cmap = plt.get_cmap('gray'))
92 plt.title('Marr-Hildreth Threshold = 15% Max(LoGImage)')
93
94 plt.subplot(2, 2, 4)
95 plt.imshow(cannyImage, cmap = plt.get_cmap('gray'))
96 plt.title('Canny')
97
98 plt.show()
99
100 originalImage = np.array(Image.open('./resource/polymersomes.tif'))
101 plt.subplot(1, 3, 1)
102 plt.imshow(originalImage, cmap = plt.get_cmap('gray'))
103 plt.title('Original')
104
105 plt.subplot(1, 3, 2)
106 globalSegmentImage = thresholdSegmentAlgorithm.globalThresholdSegment(originalImage)
107 plt.imshow(globalSegmentImage, cmap = plt.get_cmap('gray'))
108 plt.title('Global Threshold Segment')
109
110 plt.subplot(1, 3, 3)

```

```

111 otsuSegmentImage = thresholdSegmentAlgorithm.otsuThresholdSegment(originalImage)
112 plt.imshow(otsuSegmentImage, cmap = plt.get_cmap('gray'))
113 plt.title('Otsu Threshold Segment')
114
115 plt.show()

```

- 图像预处理，添加一层边缘

```

1 def imagePreprocessing(input):
2     row, col = input.shape
3     output = np.zeros((row + 2, col + 2))
4     for i in range(1, row + 1):
5         for j in range(1, col + 1):
6             output[i, j] = input[i - 1, j - 1]
7     for i in range(row + 2):
8         output[i, 0] = output[i, 1]
9         output[i, col + 1] = output[i, col]
10    for j in range(col + 2):
11        output[0, j] = output[1, j]
12        output[row + 1, j] = output[row, j]
13    return output

```

- 图像各个方向平移，以便进行求偏导的计算

```

1 def upperShift(originalImage):
2     U = np.zeros(originalImage.shape)
3     U[:, :U.shape[1] - 1] = originalImage[:, :originalImage.shape[1], :]
4     U[:, U.shape[1] - 1, :] = U[:, U.shape[1] - 2, :]
5     return U
6
7 def downShift(originalImage):
8     D = np.zeros(originalImage.shape)
9     D[:, :D.shape[1]] = originalImage[:, :originalImage.shape[1] - 1, :]
10    D[:, 0, :] = D[:, 1, :]
11    return D
12
13 def leftShift(originalImage):
14     L = np.zeros(originalImage.shape)
15     L[:, :, L.shape[2] - 1] = originalImage[:, :, 1 : originalImage.shape[2]]
16     L[:, :, L.shape[2] - 1] = L[:, :, L.shape[2] - 2]
17     return L
18
19 def rightShift(originalImage):
20     R = np.zeros(originalImage.shape)
21     R[:, :, 1] = originalImage[:, :, originalImage.shape[2] - 1]
22     R[:, :, 0] = R[:, :, 1]
23     return R

```

- Roberts, Prewitt, Sobel 算法

```

1 def Roberts(originalImage):
2     U = upperShift(originalImage)
3     L = leftShift(originalImage)
4     UL = leftShift(U)
5
6     gx = np.abs(UL - originalImage)
7     gy = np.abs(U - L)
8     M = gx + gy
9
10    return gx, gy, M
11

```

```

12 def Prewitt(originalImage):
13     U = upperShift(originalImage)
14     D = downShift(originalImage)
15     L = leftShift(originalImage)
16     R = rightShift(originalImage)
17
18     UL = leftShift(U)
19     UR = rightShift(U)
20     DL = leftShift(D)
21     DR = rightShift(D)
22
23     gx = np.abs(UR + U + UL - DR - D - DL)
24     gy = np.abs(DL + L + UL - DR - R - UR)
25     M = gx + gy
26
27     return gx, gy, M
28
29 def Sobel(originalImage):
30     U = upperShift(originalImage)
31     D = downShift(originalImage)
32     L = leftShift(originalImage)
33     R = rightShift(originalImage)
34
35     UL = leftShift(U)
36     UR = rightShift(U)
37     DL = leftShift(D)
38     DR = rightShift(D)
39
40     gx = np.abs(UR + 2 * U + UL - DR - 2 * D - DL)
41     gy = np.abs(DL + 2 * L + UL - DR - 2 * R - UR)
42     M = gx + gy
43
44     return gx, gy, M

```

- MH 算法

```

1 def LoG(x, y, sigma):
2     t = np.power(x, 2) + np.power(y, 2)
3     res = (t - np.power(sigma, 2) * 1.0) / np.power(sigma, 4) * np.exp(- t / (2 * np.power(sigma, 2)))
4     return res
5
6 def LoGThreshold(inputImage, threshold):
7     m, n = inputImage.shape
8     outputImage = inputImage.copy()
9     for i in range(1, m - 1):
10         for j in range(1, n - 1):
11             tmpList = []
12             tmpList.append(np.sum(inputImage[i - 1 : i, j - 1 : j]))
13             tmpList.append(np.sum(inputImage[i - 1 : i, j : j + 1]))
14             tmpList.append(np.sum(inputImage[i : i + 1, j - 1 : j]))
15             tmpList.append(np.sum(inputImage[i : i + 1, j : j + 1]))
16             maxValue = max(tmpList)
17             minValue = min(tmpList)
18             if maxValue > threshold and minValue < threshold:
19                 outputImage[i, j] = 255
20
21     return outputImage
22
23 def MarrHildreth(originalImage):
24     maxPix = np.max(originalImage)

```

```

25     diff = maxPix - minPix
26     originalImage = (originalImage - minPix) * 1.0 / diff
27     n = 25
28     sigma = 4
29     w = np.zeros((n, n))
30     center = (n - 1) / 2
31     for i in range(1, n + 1):
32         for j in range(1, n + 1):
33             y = i - center
34             x = j - center
35             w[i - 1, j - 1] = LoG(x, y, sigma)
36     w = w * 1.0 / np.sum(w)
37     LoGImage = originalImage - signal.convolve2d(originalImage, w, mode = 'same')
38
39     marrHildrethImage1 = LoGThreshold(LoGImage, 0)
40     threshold = 0.15 * np.max(LoGImage)
41     marrHildrethImage2 = LoGThreshold(LoGImage, threshold)
42     return LoGImage, marrHildrethImage1, marrHildrethImage2

```

- Canny 算法

```

1 def canny(img):
2     sigma = 4
3     imgdata = numpy.array(img, dtype = float)
4     gaussianImage = ndi.filters.gaussian_filter(imgdata, sigma)
5
6     emptyImage = Image.new('L', img.size)
7     gradientX = numpy.array(emptyImage, dtype = float)
8     gradientY = numpy.array(emptyImage, dtype = float)
9
10    sobelX = [[-1, 0, 1],
11               [-2, 0, 2],
12               [-1, 0, 1]]
13    sobelY = [[-1, -2, -1],
14               [0, 0, 0],
15               [1, 2, 1]]
16
17    m = img.size[1]
18    n = img.size[0]
19
20    for x in range(1, m - 1):
21        for y in range(1, n - 1):
22            px = (sobelX[0][0] * gaussianImage[x - 1][y - 1]) + (sobelX[0]
23 [1] * gaussianImage[x][y - 1]) + \
24             (sobelX[0][2] * gaussianImage[x + 1][y - 1]) + (sobelX[1]
25 [0] * gaussianImage[x - 1][y]) + \
26             (sobelX[1][1] * gaussianImage[x][y]) + (sobelX[1]
27 [2] * gaussianImage[x + 1][y]) + \
28             (sobelX[2][0] * gaussianImage[x - 1][y + 1]) + (sobelX[2]
29 [1] * gaussianImage[x][y + 1]) + \
30             (sobelX[2][2] * gaussianImage[x + 1][y + 1])
31
32            py = (sobelY[0][0] * gaussianImage[x - 1][y - 1]) + (sobelY[0]
33 [1] * gaussianImage[x][y - 1]) + \
34             (sobelY[0][2] * gaussianImage[x + 1][y - 1]) + (sobelY[1]
35 [0] * gaussianImage[x - 1][y]) + \
36             (sobelY[1][1] * gaussianImage[x][y]) + (sobelY[1]
37 [2] * gaussianImage[x + 1][y]) + \
38             (sobelY[2][0] * gaussianImage[x - 1][y + 1]) + (sobelY[2]
39 [1] * gaussianImage[x][y + 1]) + \
40             (sobelY[2][2] * gaussianImage[x + 1][y + 1])

```

```

33     gradientX[x][y] = px
34     gradientY[x][y] = py
35
36     gradientImageMag = sp.hypot(gradientX, gradientY)
37     gradientImageDir = sp.arctan2(gradientY, gradientX)
38
39     for x in range(m):
40         for y in range(n):
41             if (gradientImageDir[x][y] < 22.5 and gradientImageDir[x][y] >= 0) or \
42                 (gradientImageDir[x][y] >= 157.5 and gradientImageDir[x][y] < 202.5) or \
43                 (gradientImageDir[x][y] >= 337.5 and gradientImageDir[x][y] <= 360):
44                 gradientImageDir[x][y] = 0
45             elif (gradientImageDir[x][y] >= 22.5 and gradientImageDir[x][y] < 67.5) or \
46                 (gradientImageDir[x][y] >= 202.5 and gradientImageDir[x][y] < 247.5):
47                 gradientImageDir[x][y] = 45
48             elif (gradientImageDir[x][y] >= 67.5 and gradientImageDir[x][y] < 112.5)or \
49                 (gradientImageDir[x][y] >= 247.5 and gradientImageDir[x][y] < 292.5):
50                 gradientImageDir[x][y] = 90
51             else:
52                 gradientImageDir[x][y] = 135
53
54
55     magSup = gradientImageMag.copy()
56
57     for x in range(1, m - 1):
58         for y in range(1, n - 1):
59             if gradientImageDir[x][y] == 0:
60                 if (gradientImageMag[x][y] <= gradientImageMag[x][y + 1]) or \
61                     (gradientImageMag[x][y] <= gradientImageMag[x][y - 1]):
62                     magSup[x][y] = 0
63             elif gradientImageDir[x][y] == 45:
64                 if (gradientImageMag[x][y] <= gradientImageMag[x - 1][y + 1]) or \
65                     (gradientImageMag[x][y] <= gradientImageMag[x + 1][y - 1]):
66                     magSup[x][y] = 0
67             elif gradientImageDir[x][y] == 90:
68                 if (gradientImageMag[x][y] <= gradientImageMag[x + 1][y]) or \
69                     (gradientImageMag[x][y] <= gradientImageMag[x - 1][y]):
70                     magSup[x][y] = 0
71             else:
72                 if (gradientImageMag[x][y] <= gradientImageMag[x + 1][y + 1]) or \
73                     (gradientImageMag[x][y] <= gradientImageMag[x - 1][y - 1]):
74                     magSup[x][y] = 0
75
76     maxGrayValue = numpy.max(magSup)
77     th = 0.1 * maxGrayValue
78     tl = 0.04 * maxGrayValue
79
80
81     gnh = numpy.zeros((m, n))
82     gnl = numpy.zeros((m, n))
83
84     for x in range(m):
85         for y in range(n):
86             if magSup[x][y] >= th:
87                 gnh[x][y] = magSup[x][y]
88             if magSup[x][y] >= tl:
89                 gnl[x][y] = magSup[x][y]
90
91     gnl = gnl - gnh
92
93     def traverse(i, j):

```

```

94     x = [-1, 0, 1, -1, 1, -1, 0, 1]
95     y = [-1, -1, -1, 0, 0, 1, 1, 1]
96     for k in range(8):
97         if gnh[i + x[k]][j + y[k]] == 0 and gnl[i + x[k]][j + y[k]] != 0:
98             gnh[i + x[k]][j + y[k]] = 1
99             traverse(i + x[k], j + y[k])
100
101    for i in range(1, m - 1):
102        for j in range(1, n - 1):
103            if gnh[i][j]:
104                gnh[i][j] = 1
105                traverse(i, j)
106
107    return gaussianImage, gnh

```

- 全局閾值算法

```

1 def globalThreshold(originalImage, deltLimit = 0):
2     threshold = np.mean(originalImage)
3     deltT = threshold
4     while(deltT > deltLimit):
5         g1 = originalImage[originalImage > threshold]
6         g2 = originalImage[originalImage <= threshold]
7
8         t1 = np.mean(g1)
9         t2 = np.mean(g2)
10
11        newThreshold = (t1 + t2) / 2
12        deltT = np.abs(newThreshold - threshold)
13        threshold = newThreshold
14
15    return threshold
16
17 def globalThresholdSegment(originalImage):
18     threshold = globalThreshold(originalImage)
19     globalSegmentImage = originalImage.copy()
20     globalSegmentImage[globalSegmentImage > threshold] = 255
21     globalSegmentImage[globalSegmentImage <= threshold] = 0
22     return globalSegmentImage

```

- Ostu 閾值法

```

1 def getSigma(image, i):
2     pNumH = np.sum(image.histogram()[: i + 1])
3     pNumL = np.sum(image.histogram()[i + 1 :])
4
5     mSumH = 0
6     for j in range(1, i + 1):
7         mSumH += j * image.histogram()[j]
8
9     mSumL = 0
10    for j in range(i + 1, 255):
11        mSumL += j * image.histogram()[j]
12    try:
13        u0 = 1.0 * mSumH / pNumH
14        u1 = 1.0 * mSumL / pNumL
15        w0 = 1.0 * pNumH / (pNumH + pNumL)
16    except:
17        return 0
18
19    w1 = 1.0 - w0

```

```

20     u = (u0 - u1) ** 2
21     sigma = w0 * w1 * u
22
23     return sigma
24
25 def otsuThreshold(inputImage):
26     threshold = 0
27     gSigma = 0
28     for i in range(1,255):
29         sigma = getSigma(inputImage,i)
30         if gSigma < sigma:
31             gSigma = sigma
32             threshold = i
33     return threshold
34
35 def otsuThresholdSegment(originalImage):
36     inputImage = Image.open('./resource/polymersomes.tif').convert("L")
37     threshold = otsuThreshold(inputImage)
38     otsuSegmentImage = originalImage.copy()
39     otsuSegmentImage[otsuSegmentImage > threshold] = 255
40     otsuSegmentImage[otsuSegmentImage <= threshold] = 0
41     return otsuSegmentImage

```

实验结果

下列图片为不同边缘检测算法下的实验结果：







