

第七章作业 王晨曦

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得分：

声明：

由于考虑到手算过于冗余，故本人编写程序对数据进行统一处理，本作业中所有程序均为本人根据上课所学理论知识自己编写，无任何上网copy行为

- 给定一幅图像如下所示：

0	0	0	127	255	127	0	0	0
0	0	0	127	255	127	0	0	0
0	0	0	127	255	127	0	0	0
127	127	127	127	255	127	127	127	127
255	255	255	255	255	255	255	255	255
127	127	127	127	255	127	127	127	127
0	0	0	127	255	127	0	0	0
0	0	0	127	255	127	0	0	0
0	0	0	127	255	127	0	0	0

利用 Sobel 算子，如下所示，对它进行模板运算，

$$\text{Sobel} = \begin{bmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

要求模板运算后的图像尺寸和变换前的一致，当模板运算超出原始图像部分则对边缘进行扩展，采用 0 值对超出边界的部分进行扩展

1. 显然该Sobel是在计算y方向的梯度
2. 求取x方向梯度的模板：

-1	0	1
-1	0	1
-1	0	1

3. 求x和y方向的梯度值

- 导入图像矩阵，并将边缘填充为0

```
[[ 0.  0.  0.  0.  0.  0.  0.  0.  0.  0.  0.]
 [ 0.  0.  0.  0. 127. 255. 127.  0.  0.  0.  0.]
 [ 0.  0.  0.  0. 127. 255. 127.  0.  0.  0.  0.]
 [ 0.  0.  0.  0. 127. 255. 127.  0.  0.  0.  0.]
 [ 0. 127. 127. 127. 127. 255. 127. 127. 127. 127.  0.]
 [ 0. 255. 255. 255. 255. 255. 255. 255. 255. 255.  0.]
 [ 0. 127. 127. 127. 127. 255. 127. 127. 127. 127.  0.]
 [ 0.  0.  0.  0. 127. 255. 127.  0.  0.  0.  0.]
 [ 0.  0.  0.  0. 127. 255. 127.  0.  0.  0.  0.]
 [ 0.  0.  0.  0. 127. 255. 127.  0.  0.  0.  0.]
 [ 0.  0.  0.  0.  0.  0.  0.  0.  0.  0.  0.]]
```

- 分别将求x和y方向导数的模板写成函数sobel_x和sobel_y：

```
: def sobel_y(data, x, y):
    count = 0
    count = -data[x-1, y-1]-data[x-1, y]-data[x-1, y+1]+data[x+1, y-1]+data[x+1, y]+data[x+1, y+1]
    return count

: def sobel_x(data, x, y):
    count = 0
    count = -data[x-1, y-1]-data[x, y-1]-data[x+1, y-1]+data[x-1, y+1]+data[x, y+1]+data[x+1, y+1]
    return count
```

- 对图像矩阵使用sobel算子，并用gradient_x和gradient_y来分别存储所求得的梯度矩阵：
 - x方向梯度值：

```

]: gradient_x = np.zeros((9,9))
   for x in range(1,10):
       for y in range(1,10):
           gradient_x[x-1,y-1] = sobel_x(data, x, y)
   print(gradient_x)

```

```

[[ 0.  0. 254. 510.  0. -510. -254.  0.  0.]
 [ 0.  0. 381. 765.  0. -765. -381.  0.  0.]
 [127.  0. 254. 638.  0. -638. -254.  0. -127.]
 [382.  0. 127. 383.  0. -383. -127.  0. -382.]
 [509.  0.  0. 256.  0. -256.  0.  0. -509.]
 [382.  0. 127. 383.  0. -383. -127.  0. -382.]
 [127.  0. 254. 638.  0. -638. -254.  0. -127.]
 [ 0.  0. 381. 765.  0. -765. -381.  0.  0.]
 [ 0.  0. 254. 510.  0. -510. -254.  0.  0.]]

```

◦ y方向梯度值：

```

[12]: gradient_y = np.zeros((9,9))
      for x in range(1,10):
          for y in range(1,10):
              gradient_y[x-1,y-1] = sobel_y(data, x, y)
      print(gradient_y)

```

```

[[ 0.  0. 127. 382. 509. 382. 127.  0.  0.]
 [ 0.  0.  0.  0.  0.  0.  0.  0.  0.]
 [254. 381. 254. 127.  0. 127. 254. 381. 254.]
 [510. 765. 638. 383. 256. 383. 638. 765. 510.]
 [ 0.  0.  0.  0.  0.  0.  0.  0.  0.]
 [-510. -765. -638. -383. -256. -383. -638. -765. -510.]
 [-254. -381. -254. -127.  0. -127. -254. -381. -254.]
 [ 0.  0.  0.  0.  0.  0.  0.  0.  0.]
 [ 0.  0. -127. -382. -509. -382. -127.  0.  0.]]

```

4. 求解梯度的幅度值和方向

- 写出求解幅度值和方向的函数g_value和direction:

```
In [131]: def g_value(x,y):
count = int(np.sqrt((gradient_x[x,y]*gradient_x[x,y])+(gradient_y[x,y]*gradient_y[x,y])))
return count

In [135]: def direction(x,y):
angle0 = math.atan2(gradient_y[x,y], gradient_x[x,y])
#angle1 = int(angle0 * 180/math.pi)
return angle0
```

- 梯度的幅度值：

```
return angle0

In [78]: value = np.zeros((9,9))
for x in range(0,9):
    for y in range(0,9):
        value[x,y] = g_value(x,y)
print(value)

[[ 0.  0. 283. 637. 509. 637. 283.  0.  0.]
 [ 0.  0. 381. 765.  0. 765. 381.  0.  0.]
 [283. 381. 359. 650.  0. 650. 359. 381. 283.]
 [637. 765. 650. 541. 256. 541. 650. 765. 637.]
 [509.  0.  0. 256.  0. 256.  0.  0. 509.]
 [637. 765. 650. 541. 256. 541. 650. 765. 637.]
 [283. 381. 359. 650.  0. 650. 359. 381. 283.]
 [ 0.  0. 381. 765.  0. 765. 381.  0.  0.]
 [ 0.  0. 283. 637. 509. 637. 283.  0.  0.]
```

- 梯度的方向（弧度制）：

```
In [137]: direction_matrix = np.zeros((9,9))
for x in range(0,9):
    for y in range(0,9):
        direction_matrix[x,y] = round(direction(x,y),3)
print(direction_matrix)

[[ 0.  0.  0.464  0.643  1.571  2.499  2.678  0.  0. ]
 [ 0.  0.  0.  0.  0.  3.142  3.142  0.  0. ]
 [ 1.107  1.571  0.785  0.196  0.  2.945  2.356  1.571  2.034]
 [ 0.928  1.571  1.374  0.785  1.571  2.356  1.767  1.571  2.214]
 [ 0.  0.  0.  0.  0.  3.142  0.  0.  3.142]
 [-0.928 -1.571 -1.374 -0.785 -1.571 -2.356 -1.767 -1.571 -2.214]
 [-1.107 -1.571 -0.785 -0.196  0. -2.945 -2.356 -1.571 -2.034]
 [ 0.  0.  0.  0.  0.  3.142  3.142  0.  0. ]
 [ 0.  0. -0.464 -0.643 -1.571 -2.499 -2.678  0.  0. ]]
```

5. 求离散化后的梯度方向

- 写出离散化的函数discretization：

```
[147]: def pre_discretization(x, y):
        count = 0
        count = direction_matrix[x, y]
        if count < 0:
            count = count + np.pi
        if count > 7 * np.pi / 8:
            count = np.pi - count
        return count
```

```
[148]: def discretization(count):
        if count >= 0 and count < np.pi / 8:
            count = 0
        elif count >= np.pi / 8 and count < 3 * np.pi / 8:
            count = np.pi / 4
        elif count >= 3 * np.pi / 8 and count < 5 * np.pi / 8:
            count = np.pi / 2
        elif count >= 5 * np.pi / 8 and count <= 7 * np.pi / 8:
            count = 3 * np.pi / 4
        return count
```

- 离散化后的梯度方向（弧度制）：

```
[149]: discretization_matrix = np.zeros((9, 9))
        temp = 0
        for x in range(0, 9):
            for y in range(0, 9):
                temp = pre_discretization(x, y)
                discretization_matrix[x, y] = round(discretization(temp), 3)
        print(discretization_matrix)
```

```
[[ 0.    0.    0.785 0.785 1.571 2.356 2.356 0.    0.   ]
 [ 0.    0.    0.    0.    0.   -0.   -0.    0.    0.   ]
 [ 0.785 1.571 0.785 0.    0.    0.    2.356 1.571 2.356]
 [ 0.785 1.571 1.571 0.785 1.571 2.356 1.571 1.571 2.356]
 [ 0.    0.    0.    0.    0.   -0.    0.    0.   -0.   ]
 [ 2.356 1.571 1.571 2.356 1.571 0.785 1.571 1.571 0.785]
 [ 2.356 1.571 2.356 0.    0.    0.    0.785 1.571 0.785]
 [ 0.    0.    0.    0.    0.   -0.   -0.    0.    0.   ]
 [ 0.    0.    2.356 2.356 1.571 0.785 0.785 0.    0.   ]]
```

6. 非极大值抑制

- 非极大值抑制算法：

```
46]: result = np.zeros((9,9))
value_extent = np.zeros((11,11))
for i in range(1,10):
    for j in range(1,10):
        value_extent[i,j] = value[i-1,j-1]
for x in range(1,10):
    for y in range(1,10):
        for y in range(1,10):
            if discretization_matrix[x-1,y-1] == 0:
                if value_extent[x,y-1] < value_extent[x,y] and value_extent[x,y+1] < value_extent[x,y]:
                    result[x-1,y-1] = data[x,y]
                else:
                    result[x-1,y-1] = 0
            elif discretization_matrix[x-1,y-1] == 0.785:
                if value_extent[x-1,y+1] < value_extent[x,y] and value_extent[x+1,y-1] < value_extent[x,y]:
                    result[x-1,y-1] = data[x,y]
                else:
                    result[x-1,y-1] = 0
            elif discretization_matrix[x-1,y-1] == 1.571:
                if value_extent[x-1,y] < value_extent[x,y] and value_extent[x+1,y] < value_extent[x,y]:
                    result[x-1,y-1] = data[x,y]
                else:
                    result[x-1,y-1] = 0
            elif discretization_matrix[x-1,y-1] == 2.356:
                if value_extent[x-1,y-1] < value_extent[x,y] and value_extent[x+1,y+1] < value_extent[x,y]:
                    result[x-1,y-1] = data[x,y]
                else:
                    result[x-1,y-1] = 0
print(result)
```

算法思想：遍历图中每个像素点，根据每个点离散后的梯度值，分别与其八邻域中的两个点进行比较，若该点大于其他两个点，保留该点像素值，否则置为0。

- 结果

```
[[ 0.  0.  0. 127. 255. 127.  0.  0.  0.]
 [ 0.  0.  0. 127.  0. 127.  0.  0.  0.]
 [ 0.  0.  0. 127.  0. 127.  0.  0.  0.]
 [127. 127. 127. 127. 255. 127. 127. 127. 127.]
 [255.  0.  0. 255.  0. 255.  0.  0. 255.]
 [127. 127. 127. 127. 255. 127. 127. 127. 127.]
 [ 0.  0.  0. 127.  0. 127.  0.  0.  0.]
 [ 0.  0.  0. 127.  0. 127.  0.  0.  0.]
 [ 0.  0.  0. 127. 255. 127.  0.  0.  0.]]
```

显而易见，这是原图像的边缘。