图像细化与骨架作业

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1. 实验要求

自选一幅灰度图像(最好是数字、文字图像,或具有细线条的图像),编程实现一下功能:

- 1. 选用一种图像二值化方法,实现图像二值化,提取前景目标;
- 2. 基于二值形态学滤波方法,二值化结果进行形态学滤波,实现小尺度噪点去除和孔洞填充;
- 3. 实践一种图像细化算法, 实现前景目标细化。

要求提交原图像、二值化图像、形态学滤波结果图像和细化图像、以及自己编写的主要源码。

2. 实验步骤

2.1 otsu

otsu是阈值化中常用的自动确定阈值的方法之一,otsu 的算法试图找到一个阈值 (t),以最小化由关系给出的加权 类内方差:

$$\sigma_w^2(t) = q_1(t)\sigma_1^2(t) + q_2(t)\sigma_2^2(t)$$

其中:

$$egin{aligned} q_1(t) &= \sum_{i=1}^t P(i) \ \& \ q_2(t) = \sum_{i=t+1}^I P(i) \ & \ \mu_1(t) = \sum_{i=1}^t rac{i P(i)}{q_1(t)} \ \& \ \mu_2(t) = \sum_{i=t+1}^I rac{i P(i)}{q_2(t)} \ & \ \sigma_1^2(t) = \sum_{i=1}^t [i - \mu_1(t)]^2 rac{P(i)}{q_1(t)} \ \& \ \sigma_1^2(t) = \sum_{i=t+1}^I [i - \mu_2(t)]^2 rac{P(i)}{q_2(t)} \end{aligned}$$

代码实现如下:

```
hist_norm = hist.ravel()/hist.sum()
Q = hist_norm.cumsum()
bins = np.arange(256)
fn_min = np.inf
thresh = -1
for i in range(1, 256):
    p1, p2 = np.hsplit(hist_norm, [i]) # probabilities
    q1, q2 = Q[i], Q[255]-Q[i] # cum sum of classes
    if q1 < 1.e-6 or q2 < 1.e-6:
        continue
    b1, b2 = np.hsplit(bins, [i]) # weights
    # finding means and variances
    m1, m2 = np.sum(p1*b1)/q1, np.sum(p2*b2)/q2
    v1, v2 = np.sum(((b1-m1)**2)*p1)/q1, np.sum(((b2-m2)**2)*p2)/q2
    # calculates the minimization function</pre>
```

```
fn = v1*q1 + v2*q2
if fn < fn_min:
    fn_min = fn
    thresh = i
# find otsu's threshold value with OpenCV function
ret, otsu = cv2.threshold(blur, 0, 255, cv2.THRESH_BINARY+cv2.THRESH_OTSU)</pre>
```

2.2 形态学滤波

腐蚀可以用集合的方式定义,即

$$X\ominus S=\{x\mid S+x\subseteq X\}$$

```
def erode(img):
    h, w = img.shape
    kernal = np.array(((0, 1, 0), (1, 0, 1), (0, 1, 0)), dtype=int)
    out = img.copy()
    tmp = out.copy()
    for x in range(1, h-1):
        for y in range(1, w-1):
            if np.sum(kernal * tmp[x-1:x+2, y-1:y+2]) < 255*4:
                out[x, y] = 0
    return out</pre>
```

膨胀用集合的方式定义如下:

$$X \oplus S = \{S + x \cap X \neq \emptyset\}$$

```
def dilate(img):
    h, w = img.shape
    kernal = np.array(((0, 1, 0), (1, 0, 1), (0, 1, 0)), dtype=int)
    out = img.copy()
    tmp = out.copy()
    for x in range(1, h-1):
        for y in range(1, w-1):
            if np.sum(kernal * tmp[x-1:x+2, y-1:y+2]) >= 255:
                out[x, y] = 255
    return out
```

2.3 图像细化

设已知目标点标记为1,背景点标记为0。边界点是指本身标记为1而其8连通邻域中至少有一个标记为0的点。算法对一幅图像的所有边界点即一个3×3区域都进行如下检验和操作:

step 1 考虑以边界点为中心的8邻域,设 p_1 为中心点,对其邻域的8个点逆时针绕中心点分别标记为 p_2,p_3,\ldots,p_9 ,其中 p_2 位于 p_1 的上方。如果 $p_1=1$ (即黑点)时,下面4个条件同时满足,则标记 p_1 为删除点(最后统一删除):

```
1. 2 \leq N(p_1) \leq 6,其中N(p_1)是p_1的非零邻点的个数;
2. S(p_1)=1,其中S(p_1)是以p_2,p_3,p_4,\ldots,p_9为序时这些点的值从0到1变化的次数;
3. p_2p_6p_8=0;
4. p_4p_6p_8=0。
```

step 2 同 step 1,仅将3中的条件改为 $p_2p_4p_6=0$,④中的条件改为 $p_2p_4p_8=0$ 。同样当对所有边界点都检验 完毕后,将所有满足条件的点删除。以上两步操作构成一次迭代。算法反复迭代,直至没有点再满足标记删除的条件,这时剩下的点就组成区域的骨架。

3. 实验结果与讨论

选取的图片如图1所示,用otsu法得到的结果如图2所示,形态学滤波之后的结果如图3,最后得到的骨架如图4。



图1. 测试图



图3. 形态学滤波



图2. otsu结果图

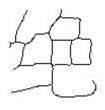




图4. 图像骨架

附录:源代码

```
import cv2
import numpy as np

# for filter

def erode(img):
    h, w = img.shape
    kernal = np.array(((0, 1, 0), (1, 0, 1), (0, 1, 0)), dtype=int)
    out = img.copy()
    tmp = out.copy()
    for x in range(1, h-1):
        for y in range(1, w-1):
            if np.sum(kernal * tmp[x-1:x+2, y-1:y+2]) < 255*4:
            out[x, y] = 0</pre>
```

```
return out
def dilate(img):
    h, w = img.shape
    kernal = np.array(((0, 1, 0), (1, 0, 1), (0, 1, 0)), dtype=int)
   out = img.copy()
   tmp = out.copy()
   for x in range(1, h-1):
       for y in range(1, w-1):
            if np.sum(kernal * tmp[x-1:x+2, y-1:y+2]) >= 255:
                out[x, y] = 255
    return out
def refine(img):
    h, w = img.shape
   out = np.zeros((h, w), dtype=int)
   out[img > 0] = 1 # chage scare to [0,1]
   out = 1-out
                        # reverse
   while True:
        delet_node1 = []
        delet node2 = []
       # step 1
       for x in range(1, h-1):
            for y in range(1, w-1):
                if out[x, y] == 1: # 如果是前景点
                    num_of_one = np.sum(out[x-1:x+2, y-1:y+2])-1 # p1的非零邻点个数
                    if num_of_one >= 2 and num_of_one <= 6: # 如果p1的非零邻点个数在2到6之间
                        if count(out, x, y) == 1: # 若从0到1的变化次数为1
                            if \operatorname{out}[x-1, y] * \operatorname{out}[x+1, y] * \operatorname{out}[x, y+1] == 0 and \operatorname{out}[x, y-1] * \operatorname{out}[x, y+1] == 0
1]*out[x+1, y]*out[x, y+1] == 0:
                                delet_node1.append((x, y)) # 将当前点加入删除点list
        for node in delet_node1: # 对删除点list进行遍历
            out[node] = 0 # 将删除点设置为0
       # step 2
        for x in range(1, h-1):
            for y in range(1, w-1):
                if out[x, y] == 1: # 如果是前景点
                    num_of_one = np.sum(out[x-1:x+2, y-1:y+2])-1 # p1的非零邻点个数
                    if num_of_one >= 2 and num_of_one <= 6: # 如果p1的非零邻点个数在2到6之间
                        if count(out, x, y) == 1: # 若从0到1的变化次数为1
                            if out[x-1, y]*out[x, y-1]*out[x+1, y] == 0 and out[x-1, y]*out[x, y-1]*out[x]
y-1]*out[x, y+1] == 0:
                                delet_node2.append((x, y)) # 将当前点加入删除点list
```

```
for node in delet_node2: # 对删除点list进行遍历
           out[node] = 0 # 将删除点设置为0
       if len(delet_node1) == 0 and len(delet_node2) == 0:
           break
   out = 1 - out # 取反
   out = out.astype(np.uint8) * 255 # 转换为0和255
    return out
def count(img, x, y):
    num = 0
   if (img[x-1, y-1] - img[x-1, y]) == 1: # p2到p3
       num += 1
   if (img[x, y-1] - img[x-1, y-1]) == 1: # p3到p4
   if (img[x+1, y-1] - img[x, y-1]) == 1: # p4到p5
       num += 1
   if (img[x+1, y] - img[x+1, y-1]) == 1: # p5到p6
       num += 1
   if (img[x+1, y+1] - img[x+1, y]) == 1: # p6到p7
       num += 1
   if (img[x, y+1] - img[x+1, y+1]) == 1: # p7到p8
       num += 1
   if (img[x-1, y+1] - img[x, y+1]) == 1: # p8到p9
       num += 1
   if (img[x-1, y] - img[x-1, y+1]) == 1: # p9到p2
       num += 1
    return num
img = cv2.imread("./image.jpg", 0)
blur = cv2.GaussianBlur(img, (5, 5), 0)
# otsu part
# find normalized_histogram, and its cumulative distribution function
hist = cv2.calcHist([blur], [0], None, [256], [0, 256])
hist_norm = hist.ravel()/hist.sum()
Q = hist_norm.cumsum()
bins = np.arange(256)
fn_min = np.inf
thresh = -1
for i in range(1, 256):
   p1, p2 = np.hsplit(hist_norm, [i]) # probabilities
   q1, q2 = Q[i], Q[255]-Q[i] # cum sum of classes
```

```
if q1 < 1.e-6 or q2 < 1.e-6:
        continue
    b1, b2 = np.hsplit(bins, [i]) # weights
   # finding means and variances
   m1, m2 = np.sum(p1*b1)/q1, np.sum(p2*b2)/q2
   v1, v2 = np.sum(((b1-m1)**2)*p1)/q1, np.sum(((b2-m2)**2)*p2)/q2
   # calculates the minimization function
   fn = v1*q1 + v2*q2
   if fn < fn_min:</pre>
        fn_min = fn
        thresh = i
# find otsu's threshold value with OpenCV function
ret, otsu = cv2.threshold(blur, 0, 255, cv2.THRESH_BINARY+cv2.THRESH_OTSU)
print("{} {}".format(thresh, ret))
# convert
result_img = img.copy()
for x in range(0, img.shape[0]):
   for y in range(0, img.shape[1]):
        if result img[x][y] < thresh:</pre>
            result_img[x][y] = 0
        else:
            result_img[x][y] = 255
# save image
# cv2.imshow("otsu result", result img)
cv2.imwrite("otsu_result.jpg", result_img)
# dilate and erode
dilate_result = dilate(result_img)
erode_result = erode(dilate_result)
erode_result = erode(erode_result)
erode_result = erode(erode_result)
dilate result = dilate(erode result)
dilate_result = dilate(dilate_result)
# cv2.imshow("res", dilate_result)
cv2.imwrite("filter_result.jpg", dilate_result)
result = refine(dilate_result)
cv2.imshow("res", result)
cv2.waitKey(0)
cv2.imwrite("refined_img.jpg", result)
```