CS270 Homework 3 Report

任怡静 2018533144

Question 1: Graph Cut for Image Segmentation (50 points)

• Please describe your algorithms in words or flowcharts. (15')

The Seed Collecting

- This part used the OpenCv built-in functions to catch mouse operations to append coordinates to seed lists (**foreground_seeds** and **background_seeds**) and illustrate seeds on image
- The image size is adjusted for drawboards so that user can draw accurately, the saved results will change the size back

• The Cut Graph Algorithm

- I first calculate a graph from seeds obtained in the previous step, which assign 0 to background_seeds' corresponding coordinates in **graph**, 1 to foreground_seeds' and 0.5 for the rest.
- Then according to the **graph**, I can construct the node list and edge list for **maxflow graph**. I append **node** (node_flatcoord, capacity_towards_source, capacity_towards_sink) into the **nodeList**, and **edge** (curr_index, neighbor_index, capacity) into the **edgeList**. It follows the rule that if the point in the **graph** is 1 then I append (node_flatcoord, 0, MAXIMUM), 0 append (node_flatcoord, MAXIMUM, 0) and (node_flatcoord, 0, 0) to the rest. I calculate and append two edges into **edgeList** for each points in the **graph**, whose capacity using $\frac{1}{1+Euclidean(I(x,y),I(x+1,y))}$ and $\frac{1}{1+Euclidean(I(x,y),I(x,y+1))}$
- Then I use built-in maxflow package to construct the maxflow graph **g**. First connect all nodes to source and sink, then for nodes which has edges to each other (have edge in **edgeList**) add edges between them. Finally do the **g.maxflow()** to get the maxflow result.
- Then I obtain the mask that the foreground is valued 1 and background 0, so that I can generate the mask and overlays based on g.get_segment(index)

• The multi-segments division

- I reuse the cut graph algorithm. T
 - To divide multiple parts, I create 4 lists to collect four kinds of seeds, select one to be the foreground and others combined to be the background
 - Loop this procedure for all four lists as they all become once the foreground. Then I check the mask before filling in color to see if it is occupied by other colors before thus only fill in the non-occupied parts.
- I modified the GUI to fit for four kinds of seeds by pressing '1', '2', '3', '4' in the keyboard
- As shown in figure 1 you need to perform graph cut method to segment the foreground and background of the given image q1_1.jpeg. (15')
 - First original image, second seed record image, third mask image, fourth mask-origin overlayed image



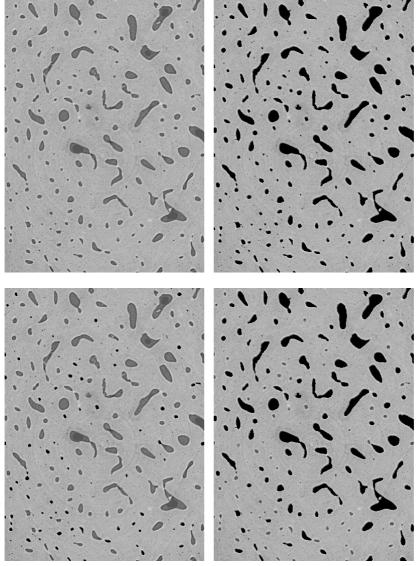
- Modify your graph cut program for multi-class segmentation. You need to segment the given image q1_2.jpeg to four parts, and show the segmentation results via transparent painting overlayed with the origin image as shown in Figure 2. (20')
 - First original image, second seed record image, third masks image, fourth masks-origin overlayed image



Question 2: Canal classification (50 points)

- Overall illustration of the process you designed. (5')
 - o Image binarize and closing, sort out background
 - In this step I use threshold() and morphologyEx() to extract the canals from the background, representing canals as white and background as black
 - **E**xtract black areas in the binarized image in the original image to form the background image **layer_background** (I_b)
 - o 8-connection white pixel finding and sort large and small canals

- \circ Calculate distortion $\sum_{i \in I_{\epsilon}} I_{\epsilon}(i)$
- Your metric of dividing canals into large and small canals. You may describe it in mathematical and natural languages. (10')
 - First I define BFS() and EightNeighbors() to find any white pixel fragments that inside
 pixels are connected by 8 connection rule in the binarized image, save their
 coordinates tuple into a list area for each fragment
 - Collect all the white fragment lists and put them into another list **segments**
 - Sort large and small canals
 - Iterate through **segments**, find element list **s** whose length is larger than 80 (contains more than 80 pixels), iterate through **s** and extract all pixels indicated by element tuples in list **s** and extract corresponding coordinates into a graph **layer_large** from the original image to gather the large canals. Vice versa, collect all small canals in another graph **layer_small**
 - layer_large = layer_large + layer_background
 - layer_small = layer_small + layer_background
- ullet Classification results, I_b , I_l and I_s . (24', 7' for each)
 - \circ First original image, second I_b , third I_l , fourth I_s



• Result of element-wised summation of $I_{\epsilon}, i.\,e.\,, \sum_{i \in I_{\epsilon}} I_{\epsilon}(i)$ (6')

 $\circ \sum_{i \in I_{\epsilon}} I_{\epsilon}(i) = 0$

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Windows PowerShell 版权所有 (C) Microsoft Corporation。保留所有权利。
营试新的跨平台 PowerShell https://aka.ms/pscore6

PS D:\Rigin_Rain\Classes\CSZ70\ShanghaitechCSZ70-DIPHws\hw3-任怡静\hw3_任怡静_2018533144> conda activate base
PS D:\Rigin_Rain\Classes\CSZ70\ShanghaitechCSZ70-DIPHws\hw3-任怡静\hw3_任怡静_2018533144> cd code
PS D:\Rigin_Rain\Classes\CSZ70\ShanghaitechCSZ70-DIPHws\hw3-任怡静\hw3_任怡静_2018533144\code> python Q2.py "..\material\images\q2.jpeg" "..\result\Q2\"
the summation of I_epsilon is: 0
PS D:\Rigin_Rain\Classes\CSZ70\ShanghaitechCSZ70-DIPHws\hw3-任怡静\hw3_任怡静_2018533144\code>
```

- Code and result illustrations, including filenames and variable names of the results above. (5')
 - For function **EightNeighbors()**:
 - It takes two parameters layer and coordinate, which layer implies the binarized 2D image, and coordinate is a tuple that means the 2D coordinate of the pixel that you want to find its 8-connected neighbors
 - The function iterate in list directions to get the pixels around, and append pixel
 that are legal (within the origin image) coordinates in tuple into the scope list
 eight_neighbors and return it
 - For function **BFSConnected()**:
 - It takes a parameter connected_list, which means the 8 connected neighbor list for the previous pixel, and initial input should be the found white pixel alone in a list.
 - The function perform a recursive BFS search from one pixel. It returns when no 8-connected neighbors are found
 - else it will create an empty list neigbor_list iterate the coordinates tuples coord in connected list:
 - Mark the pixel in the global variable layer (a copy of binarized 2D image (th3) in binarize operation) black
 - Append coord into list area, which will be new each time calls
 BFSConnected() in main function search loop (see the main function description)
 - Use EightNeighbors() to find coord's 8-connected neighbors and save them in list neighbors, iterate in neighbors, if the element is not in neibor_list created before, add it to neigbor_list, this step is to prevent redundant recursive calls since pixels in picture has more connections than a traditional
 - Finally call **BFSConnected()** on the newly obtained **neigbor_list**
 - For the main function:
 - It first read in the image, transfer it to gray image image_gray using cv2.cvtColor(), then use cv2.threshold() to binarize image_gray, save the binarized result in th3, then create two kinds of rectangle kernels of size [3,3] and [4,4], use first 3 then 4 kernel to perform cv2.morphologyEx(th3, cv2.MORPH_CLOSE, kernel) so that the noises are cleared in th3
 - Allocate two empty lists: segments and area, do a copy of th3 into layer, allocate an image-size all zero matrix layer_background, then iterate in layer:
 - If layer[i,j] > 0, meaning I find a start of an area of canal:
 - I clear up the area list, setup the input connected_list to [(i,j)] for BFSConnected(connected_list)
 - When **BFSConnected()** finished, it will append **area** to **segments**, which will save all the canal segments' coordinates lists
 - If th3[i,j] == 0, meaning I find a pixel of background

- Extract the original pixel image[i,j] into layer_background
- Allocate two image-size all zero matrix layer_large and layer_small ,lterate in segments:
 - For element list **s** in segments, if its length is larger than 80, iterate **s** and each element is (x,y), extract the original pixel **image[x,y]** into **layer_large**
 - Vice Versa, extract the original pixel **image[x,y]** into **layer_small**
- Adjust layer_large and layer_small into final results by adding layer_background into them
- Save layer_background, layer_large and layer_small
- Calculate layer_epsilon by layer_epsilon = layer_large + layer_small layer_background image, then calculate the sum of layer_epsilon by using np.sum(), print the sum result