# Computer Vision and Imaging Extended [06 30241] Assessed Assignment

- Kevin Bupalam Sudhir (2356685)

# Task 1: 2D tissue segmentation

Aim: Develop and apply different segmentation algorithms, based on any technique you have learnt to each slice of the MRI data. You need to apply exactly the same algorithm to every slice.

**Methodology:** 2D tissue segmentation can be implemented using a wide variety of image processing techniques based on the threshold, clustering, edge, region, etc. In this assignment, the image processing technique used is based on the threshold image processing technique. The algorithms used are Manual thresholding and Multi class Otsu thresholding.

**Environment used:** MATLAB R2021b **Code:** MATLAB code

# Implementation adapted for the manual and multi class Otsu thresholding:

Manual Thresholding	Multi Class Otsu thresholding
Manual thresholding algorithm manually	Multi Class Otsu thresholding is a thresholding method
calculates, and classifies the pixels in an input	that divides pixels in an input image into 'n' number of
image and can be divided into 'n' number of	classes according to the intensity of grey levels within
classes according to the intensity of grey levels	that image.
within that image.	

- 1. Load the given Brain.mat data set.
- 2. Iterate through each image label pair using for loop.
- 3. Assign an Class 1 as outer portion and Class 3 to 5 as inner portions.



- 4. Manually binarize image using mat2gray() and imbinarize() function by using grey threshold value
- 5. Then use **manual thresholding** to segment these classes.
- 4. Use multithresh() function to automatically create binary image.
- 5. Then use **Multi Class Otsu thresholding** to segment these classes.
- 6 . Get the two biggest contours and sort the connected components by area. Classes 3 to 5 has the biggest component and second biggest is the outer ring.

## Class 0 - Background class implementation

7. Manually creating background mask image from outer mask value obtained through outer ring mask and complementing it.

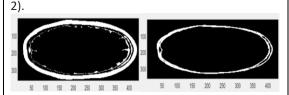


7. Use active contour to create image background mask, fill the mask with 'holes' and create component for that image.



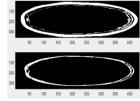
# Class 1 and 2 – Outer portion implementation

- 8. Manually find the closest value for the outer ring mask and update the same(Class 1).
- 9. From the obtained outer ring mask complement it 10. Then subtract the filled inner mask and background mask to get the inner ring mask(Class



- 8. Close all the holes in inner potion by using imfill() and extract the mask.
- 9. Using this image create another image by replacing all the pixels of value 1 (in step 8) to value 0.
- 10. Extract the mask of Class 1 and complement it.

Extract class 2 then subtract inner filled mask and background mask by using multithresh() and imquantize() to apply **multi class thresholding** and further divide image mask into 3 masks.

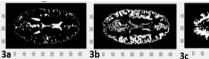


## Manual Thresholding (contd..)

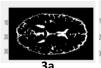
## Multi Class Otsu thresholding (contd..)

# Classes 3-5 – Inner portion implementation

- 11. Using imerode() function get the first biggest contour from the step 6.
- 12. Segment the largest contour into 3 classes by manually getting the value of the mask 3a and subtract it with background mask, outer ring mask and inner ring mask to get Class 3a. Similarly, get the value of mask 3b, then complement it and subtract it with class 3a to get class 3b.
- 13. Finally, manually get the mask 3c value from the largest contour to get class 3c (Mask 5).



- 12. Segment the largest contour into 3 classes by applying **Multi Class Otsu.**
- 13. Finally, Class 3a (Mask 3), 3b (Mask 4), 3c (Mask 5) of the biggest contour are extracted depending on the image indexes.

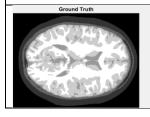




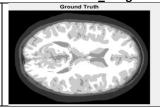


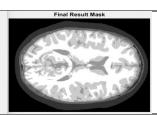
•

# 14. Final result Ground truth vs Final\_Image









Task 3: Advanced 2D tissue segmentation (3D Segmentation)

# Multi Class Otsu 3D thresholding

- 1. Load the given Brain.mat data set.
- 2. Using 3Dinput scan and **Multi Class thresholding** segment Class 1 as outer portion and Class 3 to 5 as inner portions.
- **Classes 3 to 5:** 9. Using imerode() function get the first biggest contour from the step 6.
- 10. Segment the largest contour into 3 classes by applying **Multi Class Otsu.**
- 3 . Get the two biggest contours and sort the connected components by area. Classes 3 to 5 has the biggest component and the second biggest is the outer ring.



11. Finally, Class 3a (Mask 3), 3b (Mask 4), 3c (Mask 5) of the biggest contour are extracted depending on the image indexes.

**Background Class:** 4. Use **active contour** to create background mask, fill the mask using imfill3d() function and create component for that image.









Class 1 and 2: 5. Close all the holes in inner potion by using imfill3d() and extract the 3D mask.

- 6. Using this image create another image by replacing all the pixels of value 1 (step 5) to value 0.
- 7. Apply **multi class thresholding** to further divide image mask into 3 masks and extract the mask of Class 1 and complement it.
- 8. Extract class 2 mask and then subtract inner filled mask and background mask.





12. Result: Ground truth vs Final Image Obtained







Final Image Obtained



#### Task 2: Result evaluation

The results are calculated using the following metrics:

**Structural Similarity Measure(SSIM)**: It is a measure for evaluating the visual effect of three image characteristics: brightness, contrast, and structure.

**Similarity/Intersection over Union (IOU):** It is a metric to calculate the extent of overlap between two areas. (In our case ground truth and the final label were obtained)

**Dice Score:** This is another metric that is similar to IOU where the dice coefficient is used to get the similarity of the given two samples.

**Mean Score:** This is the score that is obtained by calculating the average of the above-mentioned scores. This will give us the overall value to compare which algorithm suits the best.

#### **RESULT SCORES:**

Manual thresholding 2D					
Scores (Table 1)					
Mean Score = 0.829 Time = 2.17s					
Metric	SSIM	DICE	IOU		
M0	0.670	0.890	0.802		
M1	0.899	0.919	0.851		
M2	0.974	0.957	0.917		
M3	0.814	0.673	0.507		
M4	0.700	0.789	0.651		
M5	0.976	0.989	0.979		

Multi class Otsu thresholding 2D					
Scores (Table 2)					
Mean Score = 0.916 Time = 5.16s					
Metric	SSIM	DICE	IOU		
M0	0.987	0.997	0.994		
M1	0.960	0.977	0.939		
M2	0.955	0.922	0.845		
M3	0.879	0.849	0.725		
M4	0.888	0.958	0.896		
M5	0.936	0.979	0.931		

Multi class Otsu thresholding 3D						
Scores (Table 3)						
Mean S	Mean Score = 0.917 Time = 11.83s					
Metric	SSIM	DICE	IOU			
M0	0.987	0.997	0.995			
M1	0.948	0.972	0.945			
M2	0.954	0.916	0.846			
M3	0.859	0.851	0.742			
M4	0.874	0.945	0.897			
M5	0.892	0.963	0.930			

# **CONCLUSION:**

#### 1. Result Evaluation:

We can observe from Tables 1, 2, and 3 that the algorithms perform the best for the masks/classes 0,1,2, and 5 as these classes do not have a complex structure and also are easily differentiable. We can also observe that the mask 4 performance is a little above average, this is because the mask 4 has a lot of vacant space and is complex to get the right values. Additionally, we can observe that mask 3 performs the worst because there is a slight overlapping of class boundaries which also have a thin mask thus leading to huge error margins. Finally, we can observe that the mean values of Multi class Otsu thresholding 2D and 3D are almost the same whereas the Manual thresholding mean value is very less compared to Multi class. To summarize, Multi class Otsu thresholding is preferred over Manual thresholding.

Note: The time taken by Multiclass Otsu thresholding 2D and 3D is very high because of the active contour implementation, if active contour implementation is removed then the time taken to execute decreases significantly.

## 2. Selected algorithm for Task 3 and its reasoning

We can clearly observe from tables 1 and 2 that the performance of Manual Thresholding 2D scores per each mask/class score suffers when compared to the Multi class Otsu thresholding 2D. This behaviour is because the values of the masks in the manual thresholding are very vague and implemented manually. One more observation is that the scores suffer a lot for the masks 3 and 4 in manual thresholding as the area has a very complex structure with a lot of holes. The mean scores of the manual and multi Otsu are 0.829 and 0.916 i.e. 82.9 and 91.6 percent respectively. There is an approximate drop of 9 percent which concludes that the manual thresholding performance is bad and hence the Multi class Otsu is selected for the task 3 3D segmentation as its mean score is very good.

## 3. Comparing Multi class Otsu 2D and 3D approaches:

When the Multi class Otsu 2D is compared to Multi class Otsu 3D, from tables 2 and 3 we can see one major difference in time where the 3D implementation takes almost twice them time than 2D algorithm. This shows that 3D algorithm is slower than 2D algorithm. We can also observe that the performance of both the methods are very similar and there is a negligible difference between scores. The performance is expected to be similar as the internal functioning of algorithms are pretty similar.

# **CODES/SCRIPTS:**

Below are the codes which are used to implement the 2D and 3D segmentation. The codes are also in the following GitHub repository:

CV FinalAssignment GitHubKevin

# Code 1: manualThreshold2D.m

```
load Brain.mat
similarity_score = zeros(6,1, 'double');
ssim_array = zeros(6, 1, 'double');
dice_score = zeros(6, 1, 'double');
figure(); colormap gray; axis equal; axis off;
for i=1:10
    image = T1(:,:,i); % Reading the ith image
    1 = label(:,:,i); % Reading the ith label
    final_mask = zeros(size(1));
    image = mat2gray(image);
    level = graythresh(image);
    outer = imbinarize(image,level);
    [v, n] = bwlabel(outer, 8);
    flag = sum(bsxfun(@eq,v(:),1:n));
    [v2, n2] = maxk(flag, 2);
    t2 = n2(2);
    inval = v == t1;
    fill_imask = imfill(inval, 'holes');
    out imask = image;
    outval = fill_imask == 1;
    out_imask(outval) = 0;
    out_imask = imadjust(out_imask);
    out_rmask = out_imask > 2.352900e-01; % Value for
    outer_vals = out_rmask == 1;
    final_mask(outer_vals)=1;
    subplot(6,1,2); % Plotting mask 1
    imagesc(out_rmask);
    caption = sprintf('Mask 1');
    title(caption, 'FontSize', 8);
```

```
bg_mask = imfill(out_rmask, 'holes');
   bg_mask = imcomplement(bg_mask);
   1 values = bg mask == 1;
   final_mask(l_values)=0;
    lindex=2;
    subplot(6,1,1); % Plotting mask 0
    imagesc(bg_mask);
    caption = sprintf('Mask 0 - Background mask');
    title(caption, 'FontSize', 8);
    inner_rmask = imcomplement(out_rmask);
    inner_rmask = imsubtract(inner_rmask,fill_imask);
    inner rmask =
imsubtract(inner_rmask,double(bg_mask));
   inner_vals = inner_rmask == 1;
    final_mask(inner_vals)=lindex;
    lindex = lindex+1;
    subplot(6,1,3); % Plotting mask 2
    imagesc(inner rmask);
    caption = sprintf('Mask 2');
   title(caption, 'FontSize', 8);
    f = strel('disk', 5);
    in_imask = image;
    in vals = imerode(fill imask == 0, f);
    in_imask(in_vals) = 0;
    Xmin = min(in_imask(:));
    Xmax = max(in_imask(:));
    if isequal(Xmax,Xmin)
       in_imask = 0*in_imask;
       in_imask = (in_imask - Xmin) ./ (Xmax - Xmin);
    temp_imask = imcomplement(in_imask);
   13a_vals = temp_imask > 4.588200e-01;
img_temp = imsubtract(l3a_vals,bg_mask);
    temp_mask = zeros((size(1)));
   vals1 = img_temp==1;
    temp_mask(vals1) = 1;
   vals2 = img_temp == -1;
    temp_mask(vals2) = 0;
   13a_vals = temp_mask == 1;
    img_temp2 = imsubtract(l3a_vals,out_rmask);
    temp_mask2 = zeros((size(1)));
   vals11 = img_temp2==1;
    temp_mask2(vals11) = 1;
   vals22 = img_temp2 == -1;
    temp_mask2(vals22) = 0;
    13a_vals = temp_mask2 == 1;
    img_temp3 = imsubtract(13a_vals,inner_vals);
    temp_mask3 = zeros((size(1)));
    vals111 = img_temp3==1;
    temp_mask3(vals111) = 1;
    vals222 = img_temp3 == -1;
    temp_mask3(vals222) = 0;
    13a_vals = temp_mask3 == 1;
```

```
final_mask(13a_vals) = lindex;
    lindex = lindex+1;
    subplot(6,1,4); % Plotting mask 3
    imagesc(13a_vals);
    caption = sprintf('Mask 3');
    title(caption, 'FontSize', 8);
    % Generating mask class 3(B)
    13b_vals = in_imask > 7.137300e-01;
= imcomplement(13b_vals);
    img temp =
imsubtract(13b_vals,imcomplement(fill_imask));
    temp_mask = zeros((size(1)));
    vals1 = img_temp==1;
    temp_mask(vals1) = 1;
    vals2 = img_temp == -1;
    temp_mask(vals2) = 1;
    temp_mask =
imsubtract(temp_mask,double(13a_vals));
    13b_vals = temp_mask == 1;
    final_mask(13b_vals) = lindex;
    lindex = lindex+1;
    subplot(6,1,5); % Plotting mask 4
    imagesc(l3b_vals);
    caption = sprintf('Mask 4');
    title(caption, 'FontSize', 8);
    13c_vals = in_imask > 7.764700e-01;
final_mask(l3c_vals) = lindex;
    lindex = lindex+1;
    subplot(6,1,6); % Plotting mask 5
    imagesc(13c_vals);
    caption = sprintf('Mask 5');
    title(caption, 'FontSize', 8);
    similarity = jaccard(categorical(1),
categorical(final_mask));
    similarity_score = similarity_score + similarity;
    dice_val = dice(categorical(1),
categorical(final_mask));
    dice_score = dice_score + dice_val;
    ssim_score = get_ssim_scores(1, final_mask);
    ssim_array = ssim_array + ssim_score;
similarity_score = similarity_score / 10;
ssim_array = ssim_array / 10;
dice_score = dice_score / 10;
mean_score = (similarity_score + ssim_array +
dice_score) / 3;
meanval = mean(mean_score);
figure();colormap gray; axis equal; axis off;
imagesc(final_mask);
caption = sprintf('Final Result Mask');
title(caption, 'FontSize', 14);
% Plotting ground truth
figure();colormap gray; axis equal; axis off;
imagesc(1);
caption = sprintf('Ground Truth');
title(caption, 'FontSize', 14);
```

# Code 2: threshold2D.m

```
load Brain.mat
figure(); colormap gray; axis equal; axis off;
similarity_score = zeros(6,1,'double');
ssim_array = zeros(6, 1, 'double');
dice_score = zeros(6, 1, 'double');
for i=1:10
    image = T1(:,:,i); % Reading the ith image
    1 = label(:,:,i); % Reading the ith label
  % Creating final mask to store each segmented mask
    final_mask = zeros(size(1));
    lindex=0;
    t = multithresh(image, 1);
    outer = imquantize(image,t);
    portion mask = outer == 2;
    [v, n] = bwlabel(portion_mask, 8);
    flag = sum(bsxfun(@eq,v(:),1:n));
    [v2, n2] = maxk(flag, 2);
    t1 = n2(1);
    t2 = n2(2);
    mask = zeros(size(1));
    mask(25:end-25,25:end-25) = 1;
    img ac = activecontour(image,mask,100);
    inval = v == t1;
   mask_background = imcomplement(imfill(img_ac,
'holes'));
   f = strel('disk', 9);
    fill_imask = imclose(inval, f);
    subplot(6,1,1); % Plotting mask 0
    imagesc(mask background);
    caption = sprintf('Mask 0 - Background mask');
    title(caption, 'FontSize', 8);
    1_values = mask_background == 1;
    final mask(1 values)=lindex;
    lindex=lindex+1;
    out imask = image;
    outval = fill_imask == 1;
   out_imask(outval) = 0;
    t = multithresh(out_imask, 2); % Image mask
    v = imquantize(out imask, t);
   % Updating outer ring and outer mask to final mask
```

```
out_rmask = v == 1;
   out rmask = imcomplement(out rmask);
   outer_vals = out_rmask == 1;
   final_mask(outer_vals)=lindex;
   lindex = lindex+1;
   subplot(6,1,2); % Plotting mask 1
    imagesc(out_rmask);
   caption = sprintf('Mask 1');
   title(caption, 'FontSize', 8);
   inner rmask = zeros((size(1)));
   inner_ring_vals = v == 2;
   inner_rmask(v == 1) = 1;
   bg_vals = mask_background == 1;
   inner_rmask(bg_vals) = 0;
    inner_vals = fill_imask ==1;
   inner_rmask(inner_vals) = 0;
   inner_mask_vals = inner_rmask == 1;
   final_mask(inner_mask_vals)=lindex;
   lindex = lindex+1;
   subplot(6,1,3); % Plotting mask 2
   imagesc(inner rmask);
   caption = sprintf('Mask 2');
   title(caption, 'FontSize', 8);
   f = strel('disk', 8);
   in_imask = image;
   in vals = imerode(fill imask == 0, f);
    in_imask(in_vals) = 0;
   t = multithresh(in_imask, 3); % Into 3 classes
   v_in = imquantize(in_imask, t);
   13a_vals = v_in == 2;
   final_mask(13a_vals) = lindex;
   lindex = lindex+1;
   subplot(6,1,4); % Plotting mask 3
   imagesc(13a_vals);
   caption = sprintf('Mask 3');
   title(caption, 'FontSize', 8);
   13b_vals = v_in == 3;
   final_mask(13b_vals) = lindex;
   lindex = lindex+1;
   subplot(6,1,5); % Plotting mask 4
   imagesc(13b_vals);
   caption = sprintf('Mask 4');
   title(caption, 'FontSize', 8);
   13c_vals = v_in == 4;
   final_mask(13c_vals) = lindex;
   lindex = lindex+1;
   subplot(6,1,6); % Plotting mask 5
   imagesc(13c_vals);
   caption = sprintf('Mask 5');
   title(caption, 'FontSize', 8);
   similarity = jaccard(categorical(1),
categorical(final_mask));
    similarity_score = similarity_score + similarity;
```

```
dice_val = dice(categorical(1),
categorical(final mask));
    dice_score = dice_score + dice_val;
    ssim_score = get_ssim_scores(1, final_mask);
    ssim array = ssim array + ssim score;
similarity_score = similarity_score / 10;
ssim_array = ssim_array / 10;
dice_score = dice_score / 10;
mean score = (similarity score + ssim array +
dice_score) / 3;
mean_val = mean(mean_score);
figure();colormap gray; axis equal; axis off;
imagesc(final_mask);
caption = sprintf('Final Result Mask');
title(caption, 'FontSize', 14);
figure();colormap gray; axis equal; axis off;
imagesc(1);
caption = sprintf('Ground Truth');
title(caption, 'FontSize', 14);
```

#### Code 3: threshold3D.m

```
load Brain.mat
image = T1; % Reading the image
1 = label; % Reading the label
final_mask = zeros(size(1));
lindex = 0;
t = multithresh(image, 1);
outer = imquantize(image, t);
p_mask = zeros(size(outer));
t_vals = outer == 2;
p mask(t vals) = 1;
v = bwlabeln(p_mask); % Get components that are
in_rmask = zeros(size(outer));
invals = v == 3:
in_rmask(invals) = 1;
mask = zeros(size(1));
mask(25:end-25,25:end-25) = 1;
img_ac = activecontour(image,mask,100);
mask_background = imfill3d(img_ac);
fill_imask = imfill3d(in_rmask);
maskbg = zeros((size(1)));
1_values = mask_background == 0;
maskbg(l_values) = 1;
final_mask(l_values) = lindex;
lindex = lindex+1;
```

```
out_mask = image;
outvals = fill_imask == 1;
out mask(outvals) = 0;
t = multithresh(out_mask, 2); % Image mask conversion
v = imquantize(out_mask, t);
out_rmask = zeros((size(1)));
outval = v == 1;
out_rmask(outval) = 1;
out_rmask = imcomplement(out_rmask);
outer_rvals = out_rmask == 1;
final_mask(outer_rvals)=lindex;
lindex = lindex+1;
in_rmask = zeros((size(1)));
in_rvals = v == 2;
in_rmask(outval) = 1;
1 values = maskbg == 1;
in_rmask(l_values) = 0;
invals = fill_imask ==1;
in_rmask(invals) = 0;
in_maskvals = in_rmask == 1;
final mask(in maskvals)=lindex;
lindex = lindex+1;
f = strel('disk', 8);
in_imask = image;
invals = imdilate(fill_imask, f)==0;
in_imask(invals) = 0;
t = multithresh(in_imask, 3); % Into 3 classes
v_in = imquantize(in_imask, t);
13a_mask = zeros((size(1)));
13a_vals = v_in == 2;
13a_mask(13a_vals) = 1;
final_mask(13a_vals) = lindex;
lindex = lindex+1;
13b_mask = zeros((size(1)));
13b_vals = v_in == 3;
13b_mask(13b_vals) = 1;
final_mask(13b_vals) = lindex;
lindex = lindex+1;
13c_mask = zeros((size(1)));
13c_vals = v_in == 4;
13c_mask(13c_vals) = 1;
13c_vals = 13c_mask == 1;
final_mask(13c_vals) = lindex;
lindex = lindex+1;
similarity = jaccard(categorical(1),
categorical(final mask));
ssim_score = get_ssim_scores(1, final_mask);
dice_score = dice(categorical(1),
```

```
categorical(final_mask));
%Compute mean of all three calculated scores
mean_score = (similarity + dice_score + ssim_score) /
3;
mean_val = mean(mean_score);
% Plotting final result mask
figure();
volshow(final_mask);
figure();
volshow(1);
% Function to fill image along all the three
dimensions
function f3=imfill3d(f3)
    for i=1:3
        for j=1:size(f3,3)
            f3(:,:,j)=imfill(f3(:,:,j),'holes');
        end
end
```

# Code 4: get\_ssim\_scores.m

```
% Function to calculate ssim scores (Structural
Similarity)
function values=get_ssim_scores(grount_truth, 1)
    values = zeros(6,1,'double');
    for i=0:5
        grount_truth_mask =
zeros((size(grount_truth)));
        grount_truth_vals = grount_truth == i;
        grount_truth_mask(grount_truth_vals) = 1;

    mask = zeros((size(grount_truth)));
    vals = 1 == i;
    mask(vals) = 1;
    [ssim_values, ~] = ssim(grount_truth_mask,
mask);
    values(i+1) = values(i+1) + ssim_values;
    end
end
```