

BUILDING AN ASSURANCE CASE FOR
AORTAGEOMRECON SOFTWARE

BUILDING AN ASSURANCE CASE FOR AORTA GEOMETRY
RECONSTRUCTION SOFTWARE

BY
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TITLE: Building an Assurance Case for Aorta Geometry Reconstruction Software

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Abstract

Assurance cases has been proven to be effective developing a real-time system software.

Another domain that requires the high standard correctness, completeness is medical software.

Throughout the development of the Aorta Geometry Reconstruction software, we implicitly listed the evidences that are essential to build our confidence in the software for assurance cases, build the artifact and the evidences simultaneously.

Finally, we present this software with the list of the evidences built for assurance cases, to show that the assurance cases can apply well on the medical software

Your Dedication
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Acknowledgements

Acknowledgements go here.

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Notation, Definitions, and Abbreviations

Notation

$A \leq B$ A is less than or equal to B

Definitions

Aorta The aorta is the largest artery of the body and carries blood from the heart to the circulatory system. It has several sections: The Aortic Root is the transition point where blood first exits the heart. It functions as the water main of the body. The Aortic arch, is the curved segment that gives the aorta its cane-like shape. It bridges the ascending and descending aorta. Throughout the documentation, Aorta would only include Ascending aorta, Aortic arch and Descending aorta. Abdominal aorta is not considered as the interested part.

Ascending Aorta

The ascending aorta is the first part of the aorta, which is the largest blood vessel in the body. It comes out of your heart and pumps blood through the aortic arch and into the descending aorta.

Descending Aorta

The descending aorta is the longest part of your aorta (the largest artery in your body). It begins after your left subclavian artery branches from your aortic arch, and it extends down into your belly. The descending aorta runs from your chest (thoracic aorta) to your abdominal area (abdominal aorta).

Organ Segmentation

The definition of the organ boundary or organ segmentation is helpful for the orientation and identification of the regions of interest inside the organ during the diagnostic or treatment procedure. Further, it allows the volume estimation of the organ, such as the aorta.

Inferior Inferior is the direction away from the head; the lower (e.g., the foot is part of the inferior extremity).

Superior Superior is the direction toward the head end of the body; the upper (e.g., the hand is part of the superior extremity).

Slice A 2-dimensional image is retrieved from a 3-dimensional volume.

Kernel Size The size of the kernel for binary dilation.

Label Map A labeled map or a label image is an image that labels each pixel of a source image.

Binary Dilation

Binary dilation is a mathematical morphology operation that uses a structuring element (kernel) for expanding the shapes in an image.

rms_error Value of RMS change below which the filter should stop. This is a convergence criterion.

Maximum iteration

Number of iterations to run

Curvature scaling

Weight of the curvature contribution to the speed term.

Propagation scaling

Weight of the propagation contribution to the speed term.

Segmented slice

A 2-dimensional image retrieved by applying SITK's ThresholdSegmentationLevelSetImageFilter with the euclidean distance transform image, the original image, and the threshold value calculated with the mean and the standard deviation of the intensity values that were labeled as the white pixel.

Contour Line A contour line (also isoline, isopleth, or isarithm) of a function of two variables is a curve along which the function has a constant value so that the curve joins points of equal value.

Level Sets Level Sets are an important category of modern image segmentation techniques based on partial differential equations (PDE), i.e. progressive evaluation of the differences among neighboring pixels to find object boundaries. The pictures below demonstrate an example of how Level Sets method work on finding the region of the heart. It starts with a seed contour that is within the region of interest, then by finding the gradient based on the contour line, the segmentation result will propagate towards outside of the region until the maximum difference between the neighboring pixels are reached.

Threshold Coefficient

This coefficient is used to compute the lower and upper threshold passing through the segmentation filter SITK's `ThresholdSegmentationLevelSetImageFilter`. The algorithm first uses SITK's `LabelStatisticsImageFilter` to get the mean and the standard deviation of the intensity values of the pixels that are labeled as the white pixel. Larger values with this coefficient imply a larger range of thresholds when performing the segmentation, which leads to a larger segmented region.

Stop Limit This limit is used to stop the segmentation algorithm. It is used differently in segmentation in inferior direction and segmentation in superior direction.

Euclidean distance transform

The euclidean distance transform is the map labeling each pixel of

the image with the distance to the nearest obstacle pixel (black pixel for this project).

DICOM Digital Imaging and Communications in Medicine (DICOM) is the standard for the communication and management of medical imaging information and related data.

Abbreviations

AortaGeomRecon

3D Slicer’s extension module, Aorta Geometry Reconstruction

DICOM Digital Imaging and Communications in Medicine

MG Module Guide

SITK SimpleITK

SRS Software Requirements Specification

Declaration of Academic Achievement

The student will declare his/her research contribution and, as appropriate, those of colleagues or other contributors to the contents of the thesis.

Chapter 1

Introduction

This chapter includes an introduction for AortaGeomRecon assurance cases study. In section [1.1](#), we are discussing the objective of this case study, and this documentation. In section [1.2](#), we are explaining some terms and definitions used throughout this document. Section [1.3](#) discussed about the problem statement, and finally section [1.4](#) gives the brief outline for this document.

1.1 Objective

In this study, we present the result of applying of assurance cases in a developing medical software to build the stakeholders' confidence in this software.

1.2 Background

Aorta is the largest artery that carries blood from the heart to the circulatory system. It has a cane-like shape with Ascending aorta, Aortic arch and Descending aorta.

Unlike the real cane stick, the descending aorta might not be straight for each patient.

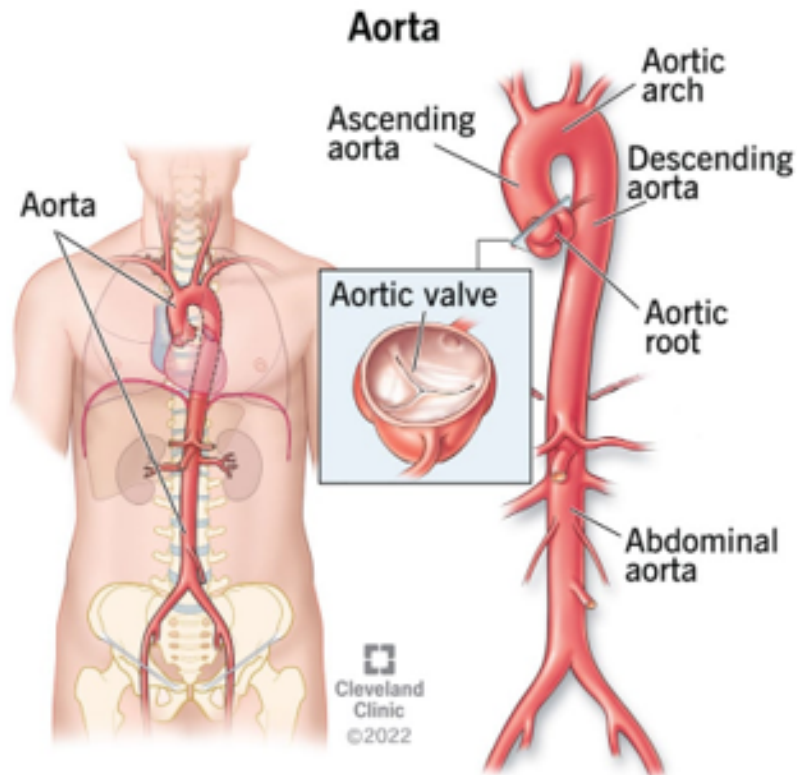


Figure 1.1: Aorta

Aorta segmentation in CT scans is important for:

- Coarctation of the aorta
- Aortic calcification quantification
- To guide the segmentation of other central vessels.

Assurance cases

3D Slicer

Image Processing

1.3 Problem Statement

The main goals of this project is building a software that can quickly build the 3D geometry of the Aorta from CT Chest scans, while applying the assurance cases for this academic medical software. This project shows the assurance cases can indeed help build up our confidence in the medial software in general, because medical software like real-time system software, needed completeness and correctness.

Build Software and Assurance cases for this software. Start with a list of Functional and Non-Functional requirements.

1.4 Thesis Outline

Chapter 2

AortaGeomRecon Research and Development

This chapter will discuss about the research and development of the AortaGeomRecon.

AortaGeomRecon stands for Aorta Geometry Reconstruction. The main objective of this software is to semi-automatically build 3D geometry of the Aorta from the patient's chest ct scans. The existing methods are often involved of extensive manual works by using a software with many steps. An experienced user, who might be a medical domain expert, needs to do a minimum of 10 minutes of manual works.

The implementation till the date of this report can let the users who have the user characteristics described in SRS ([Lin, 2023](#)) get the Aorta 3D geometry with only a few hyperparameters which can be set within half a minute, and the result requires maximum 2 minutes of execution time.

2.1 Existing Methods

There are many segmentation software available to the users, we will discuss the two builtIn methods in two softwares, ITK-nap and 3D Slicer.

2.1.1 ITK-Snap bubble method

ITK-Snap provides a segmentation method that first let user to select multiple voxels with a custom intial size and expanding size within the volume. We refer this method as “bubble method”.

Through many iterations, the voxels expand to fill the entire volume, finally user will need to cut the extra part of the volume. This Figure 2.1 shows the ITK-Snap UI executing the segmentation.

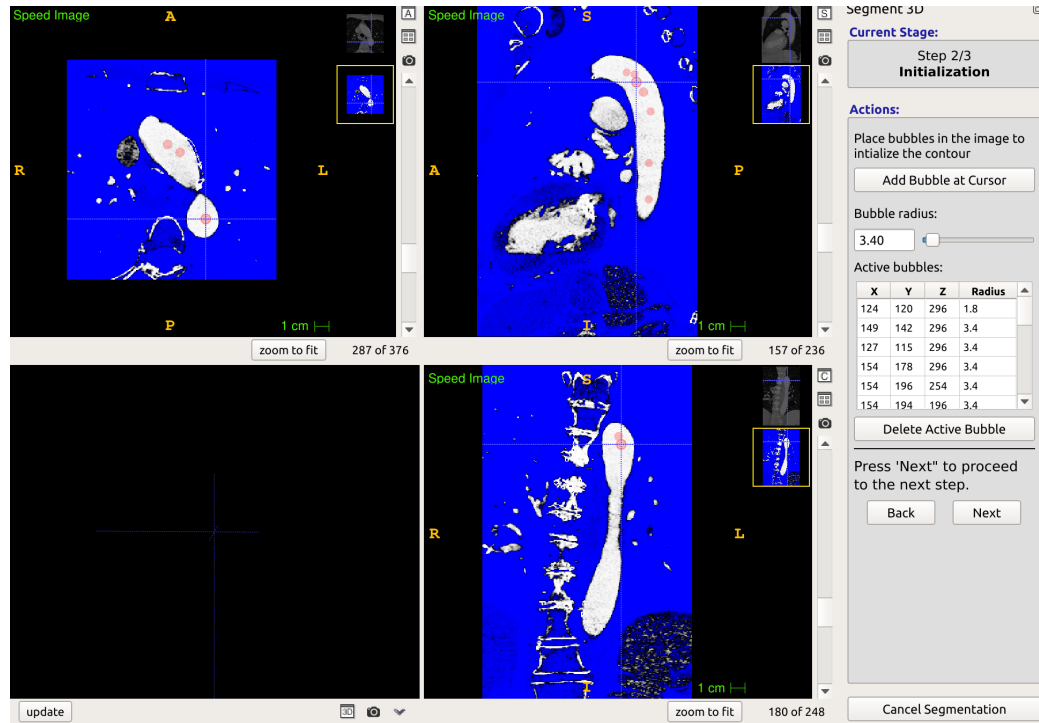


Figure 2.1: ITK-Snap’s Bubble segmentation method

The advantages of the bubble method is that it guarantes to produce a correct the segmentation result. A medical domain expert can manually control the wanted area, and visually observing the segmentation result expanding, shrinking and the user can erase the unwanted part.

The disadvantages of this method is that the operations described above are complicated. Easier to say then do, an opearator who has previous experience building the geometry with this method still needed 20 minutes of manual work building a new aorta geometry. Plus, ITK-Snap software can only read VTK file, therefore the chest CT scans that are usually DICOM, needed a manual conversion before using this software and its segmentation method.

2.1.2 3D Slicer threshold segmentation

3D Slicer is another well-known medical image processing software for academic. 3D Slicer provides multiple segmentation methods, and one of the quickest and easiest to use is the intensity based segmentation.

This method first let user select a small area that belongs to the wanted area on a 2D plane (Axial, Sagittal, and Coronal). 3D Slicer read the pixels' intensity of the surrounding area, and segment based on the intensity threshold. Any pixels's intensity that is within the range will be segmented as the segmentation result.

Like the bubble method, this method often reads extra volume, and requires user to cut the unwanted parts. A [YouTube video](#) shows an experience user who gets the aorta 3D geometry with 8 minutes of manual works.

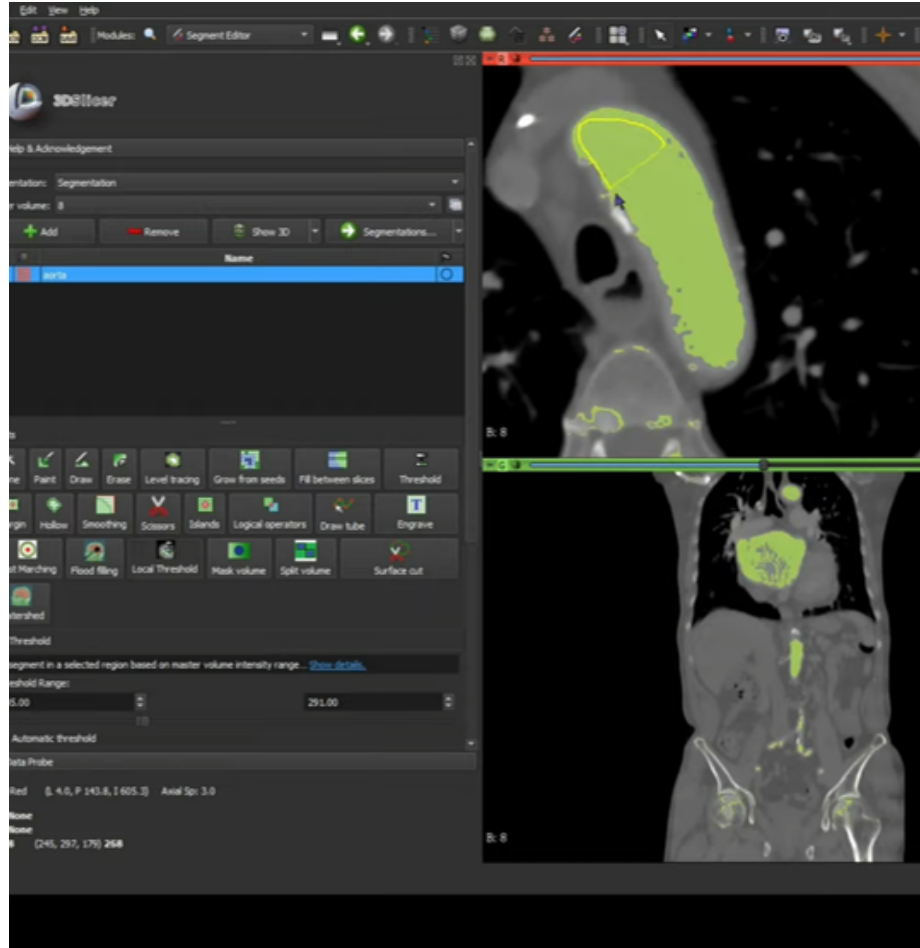


Figure 2.2: 3D Slicer BuiltIn Segmentation Method

2.2 Segmentation Algorithm

This section introduces the key concepts of the implementation on the segmentation algorithm. The algorithm is developed in Python, by using external libraries, SimpleITK and Numpy. The algorithm build the 3D Aorta geometry by doing segmentation on each axial slice. From the axial view, there is one or two circles that

is edged bouded. Using this information, and given an initial Aorta centre coordination, the algorithm continously segments each axial's slice circle closest to the previous Aorta centre coordinate. Finally, some hyperparameters tuning can let the algorithm pick up the pixels that were missing but belongs to the part of the Aorta.

2.2.1 Background

SimpleITK is an open-source multi-dimensional image analysis library Developed by the Insight Toolkit community for the biomedical sciences and beyond. NumPy is the fundamental package for scientific computing with Python, especially for the performance on multi-dimensional array processing. The algorithm will use functions from these two libraries for image processing and multi-dimensional array processing. For example, the algorithm segments each slice with `ThresholdSegmentationLevelSetImageFilter` from SITK.

The algorithm works best with the chest volume cropped to a rectangular prism that contains the aorta and parts of the other organs such as the backbone, blood vessels, and the heart. This can be done with 3D Slicer and its builtIn modules, Volume rendering and Crop Volume, or researched by the user to find the starting point and the size to crop.

2.2.2 Parameters

At the beginning of the algorithm, the user inputs two integer coordinates indicating the position of the descending aorta and ascending aorta centre on a single slice. The yellow dots in Figure 2.3 shows an example of the aorta seeds. These seeds will be updated by the algorithm after processing each axial plane.

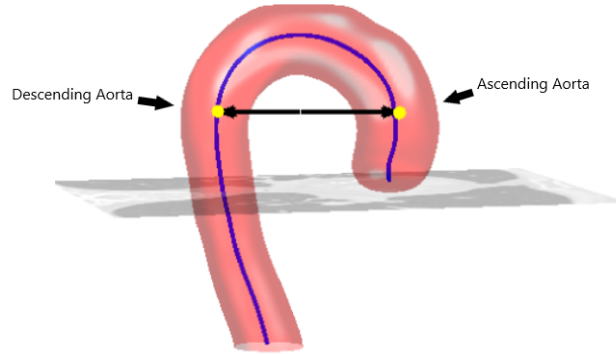


Figure 2.3: The aorta seeds ([Kurugol et al., 2012](#))

On the other hand, this list of hyperparameters that can be tuned to get the best segmentation result:

- The stop limit which controls the stop condition
- The threshold coefficient which controls the segmentation acceptable intensity range
- The kernel size which controls the label image circle size
- The threshold Segmentation Level Sets Image Filter parameters, including:
 - The rms error
 - The maximum iteration
 - The curvature scaling
 - The propagation scaling

One of the most important parameters is the threshold coefficient. Since the algorithm segments based on the intensity of the grey scale pixels, decreasing the threshold coefficient would decrease the acceptable range of the pixels, and vice-versa.

2.2.3 Algorithm Overview

When the user has selected the aorta seeds, the plane where the aorta seeds located is the initial plane. From this plane towards the bottom (the feet) is inferior direction. On the other hand, it is superior direction. This algorithm segments each slice with `SITK::ThresholdSegmentationLevelSetImageFilter`. The principles of this image filter can be explained with two terms: Level sets segmentation method, and a threshold range that defines the intensity of the acceptable pixel.

Level Sets are an important category of modern image segmentation techniques based on partial differential equations (PDE), i.e. progressive evaluation of the differences among neighboring pixels to find object boundaries. The pictures below demonstrate an example of how Level Sets method work on finding the region of the heart. It starts with a seed contour that is within the region of interest, then by finding the gradient based on the contour line, the segmentation result will propagate towards outside of the region until the maximum difference between the neighboring pixels are reached.

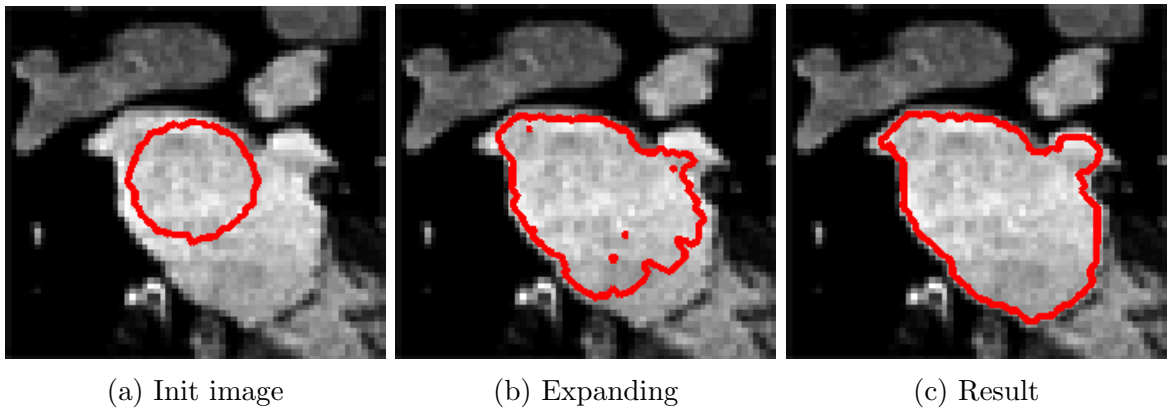


Figure 2.4: Level Sets Segmentation

2.2.4 The steps to segment a single slice

In the following section, we will present each step to segment a single slice. These steps applied in both segmentation in superior and inferior direction, there is a difference in the stop condition, which we will elaborate in the following section.

2.2.4.1 Create A Label Map

The algorithm uses `SITK::BinaryDilateImageFilter` to perform binary dilation to generate a circle-like shape around the centre coordinates (user input's for the first slice and calculated by the algorithm for the rest of the slices). Each pixel within this shape will be labeled as a white pixel (value of 1), and the rest of the pixels are labeled as black pixels (value of 0).

The generated result is the label map image, and we will use it in the next few steps. The size of the circle-like shape is determined by the kernel size (user's input). The Figure 2.5 shows an example of generated label map image (the green parts) overlay over the original slice.

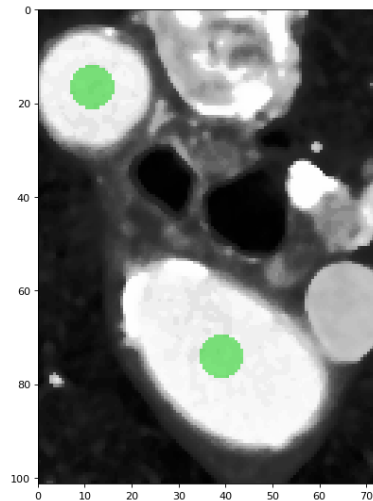


Figure 2.5: A label map

2.2.4.2 Create A Distance Map

With `SITK::SignedMaurerDistanceMapImageFilter`, the algorithm creates another image, the Euclidean distance transform of the label image from previous step. This is used as a contour line that helps build the gradient mentioned in Level sets. The Figure 2.6 shows an example of distance map.

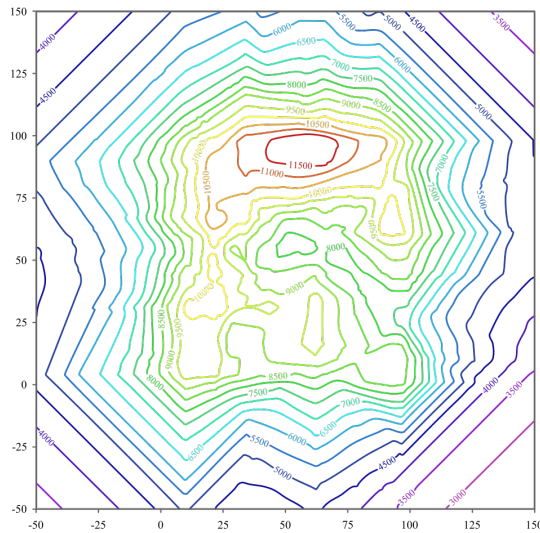


Figure 2.6: A distance map

2.2.4.3 Calculate a threshold range

By using `SITK::LabelStatisticsImageFilter`, the algorithm gets the mean and the standard deviation of the intensity values of the pixels that were labeled as the white pixel in the label map. The algorithm uses threshold coefficient to calculate the lower and upper threshold to be used in the next step.

```
intensity_mean = self._stats_filter.GetMean(  
    PixelValue.white_pixel.value)  
std = self._stats_filter.GetSigma(PixelValue.white_pixel.value)  
lower_threshold = (intensity_mean - self._threshold_coef*std)  
upper_threshold = (intensity_mean + self._threshold_coef*std)  
self._segment_filter.SetLowerThreshold(lower_threshold)  
self._segment_filter.SetUpperThreshold(upper_threshold)
```

Figure 2.7: Code snippet shows the intensity threshold calculation

2.2.4.4 Segment a single slice

With `SITK::ThresholdSegmentationLevelSetImageFilter`, the seed image calculated in step 2.2.4.2, and the lower and upper threshold value calculated in step 2.2.4.3, the algorithm performs segmentation and generated a segmented slice. The Figure 2.8 shows an example of generated segmented slice (the green parts) overlay over the original slice.

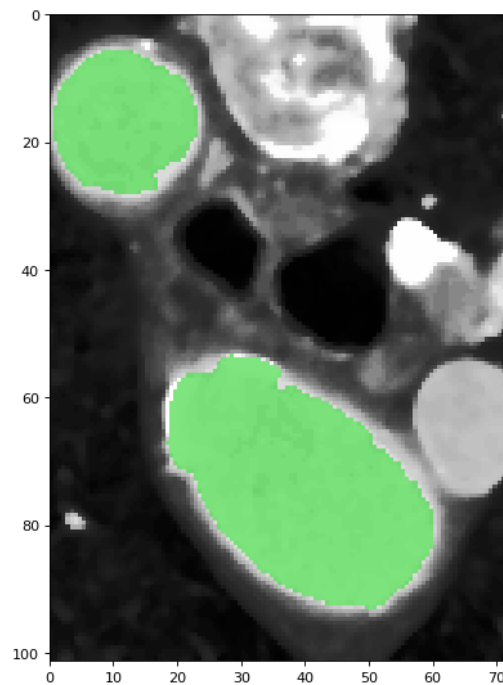


Figure 2.8: A segmented image on top of the original slice

2.2.4.5 Calculate new centroids

By comparing each pixel segmented as aorta to the previous descending centroid and the previous ascending centroid, the algorithm use the positions of the points closer to the previous descending centroid to calculate new descending aorta centroid, and vice-versa for the ascending aorta centroid. However, at certain point during the segmentation in inferior direction, the slice might reaches the end of the ascending aorta, the Aortic Root, the algorithm will stop using and calculate ascending aorta centroid and only computes descending aorta centroid for the slices afterward.

2.2.4.6 Verify segmentation result

There are two main stop conditions for verifying segmentation result, one condition for the segmentation in inferior direction and the other one for the segmentation in superior direction. Stop limit is an user defined parameter to control the algorithm, to calculate the condition with the centroids position and the standard deviation.

In inferior direction, if the new ascending aorta centroid that is closest to the previous ascending aorta centre is reaching the distance limit, then the algorithm will stop consider taking the new centroid closer to the ascending aorta. In other words, only 1 centroid will be used for descending aorta segmentation.

In superior direction, if the standard deviation of the initial label map and the segmentation result label map has larger difference than the limit, the algorithm will stop processing segmentation for the rest of the slices. For example, assume that the standard deviation of the initial label image is 20, and the standard deviation of the segmentation label image is 40, with stop limit of 10, the program halt immediately.

2.2.5 Algorithm Summary

Given two integer coordinates, ascending aorta centroid and descending aorta centroid, the algorithm set the inputed plane as the initial plane. From the initial plane to the bottom (the feet) plane, the algorithm calculate a label map with two centroid coordinates and kernel size, calculate a distance map with the label map, calculate a threshold range with the label map's selected pixels, perform segmentation, calculate new centroid coordinates, and verify the segmentation result in case that it reaches the stop condition. Once the algorithm finished the segmentation in inferior direction, the algorithm works from the initial plane to the top (the head) plane, repeating the similar steps. Each segmentation result slice is stored in SITK's image, which supports the conversion to VTK file or DICOM file.

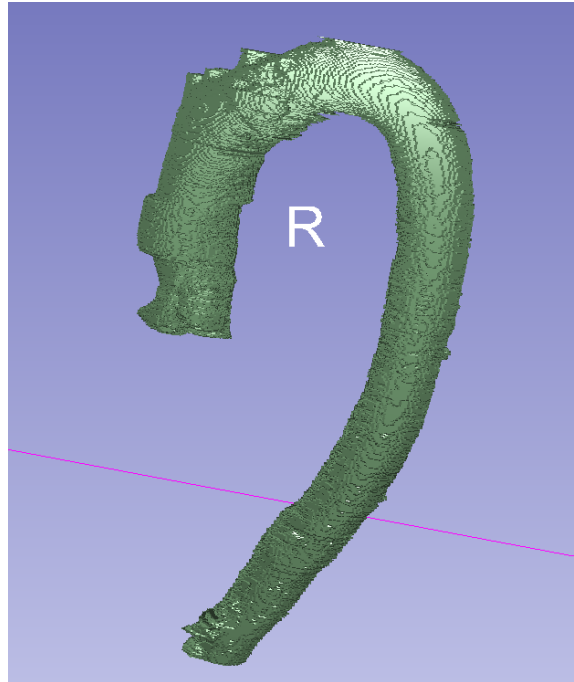
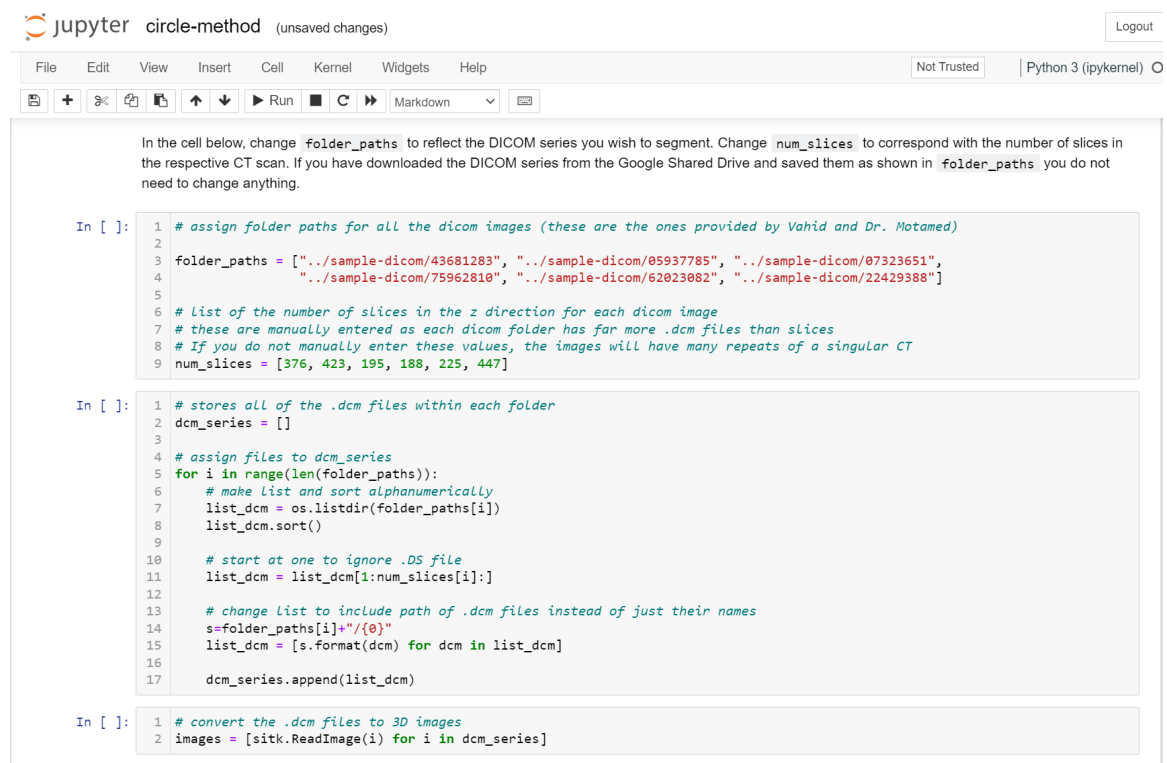


Figure 2.9: Segmentation Result

2.3 3D Slicer Extension Development

The project has started with the a simple segmentation algorithm build on the jupyter notebook. When getting a new patient’s data, the user will need to investigate the chest ct scans using another software (3D Slicer, ITK-Snap), to get the readings such as coordinates and size to crop (the coordinates of the yellow dots shown in Figure 2.3).



The screenshot shows a Jupyter Notebook titled "circle-method" with unsaved changes. The interface includes a top bar with "File", "Edit", "View", "Insert", "Cell", "Kernel", "Widgets", and "Help" menus. Below the menus is a toolbar with icons for running, saving, and other actions. The notebook content consists of three code cells. The first cell contains comments and lists of folder paths and slice counts. The second cell contains a loop that processes each folder path, listing and sorting DICOM files, and then selecting a subset based on the slice count. The third cell contains a loop that reads each DICOM file into a 3D image object.

```
In [ ]: 1 # assign folder paths for all the dicom images (these are the ones provided by Vahid and Dr. Motamed)
2
3 folder_paths = ["../sample-dicom/43681283", "../sample-dicom/05937785", "../sample-dicom/07323651",
4               "../sample-dicom/75962810", "../sample-dicom/62023082", "../sample-dicom/22429388"]
5
6 # list of the number of slices in the z direction for each dicom image
7 # these are manually entered as each dicom folder has far more .dcm files than slices
8 # If you do not manually enter these values, the images will have many repeats of a singular CT
9 num_slices = [376, 423, 195, 188, 225, 447]

In [ ]: 1 # stores all of the .dcm files within each folder
2 dcm_series = []
3
4 # assign files to dcm_series
5 for i in range(len(folder_paths)):
6     # make list and sort alphanumerically
7     list_dcm = os.listdir(folder_paths[i])
8     list_dcm.sort()
9
10    # start at one to ignore .DS file
11    list_dcm = list_dcm[1:num_slices[i]:]
12
13    # change list to include path of .dcm files instead of just their names
14    s=folder_paths[i]+"/{0}"
15    list_dcm = [s.format(dcm) for dcm in list_dcm]
16
17    dcm_series.append(list_dcm)

In [ ]: 1 # convert the .dcm files to 3D images
2 images = [sitk.ReadImage(i) for i in dcm_series]
```

Figure 2.10: Jupyter Notebook Researched by Kailin Chu

Originally, the input file paths, the total number of slices and the inputs by the users are hardcoded in the Jupyter Notebook. To improve the usability of the Aorta-GeomRecon (reduce the amount of time for user inputs and execution), we implemented an extension module on 3D Slicer.

3D Slicer is a open-sourced medical image processing software for research. 3D Slicer provides useful modules such as Crop Volume module and Volume Rendering module, and it's highly modulizable with Python Scripting to control the extension module sequence, and QT designer to generate Graphical User Interface.

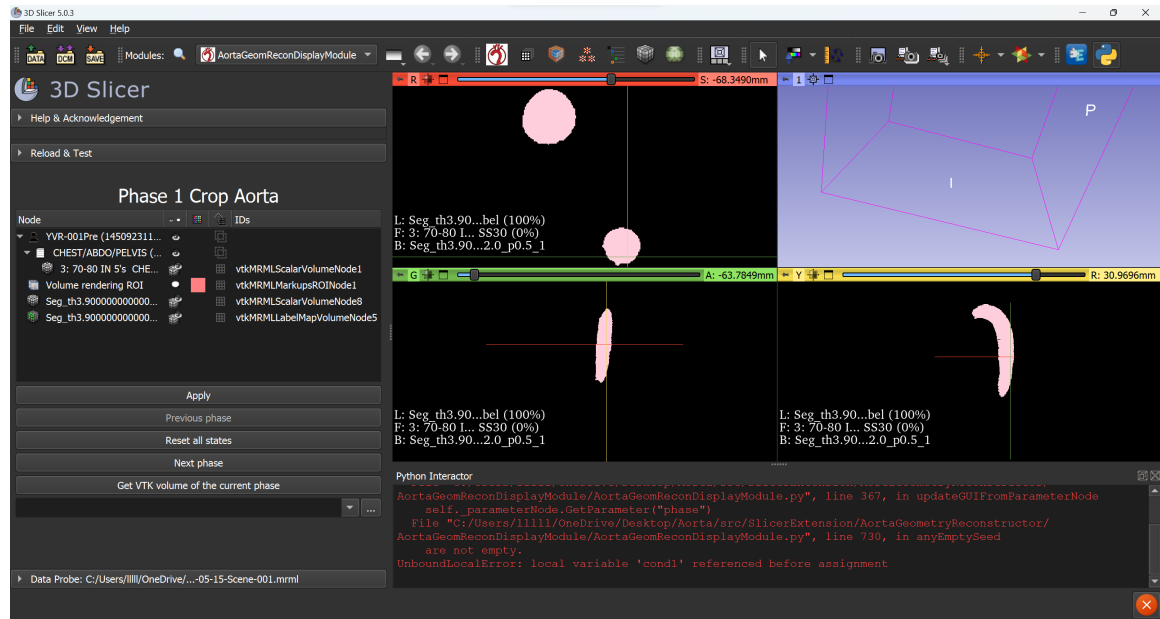


Figure 2.11: AortaGeomRecon UI phase 1

3D Slicer supports modulization with an extension. An extension is a software bundle that includes every modules to solve a specific medical image problem. An extension can compose multiple module, where each module is dedicated to solve a sub-problem.

2.3.1 3D Slicer's data structure

3D Slicer's Data Structure can be divided into two categories: Node for storing large data such as DICOM with a Volume Node, Volume rendering Region of Interest Node, Label Map Volume Node, and parameters are read from extension's module UI and

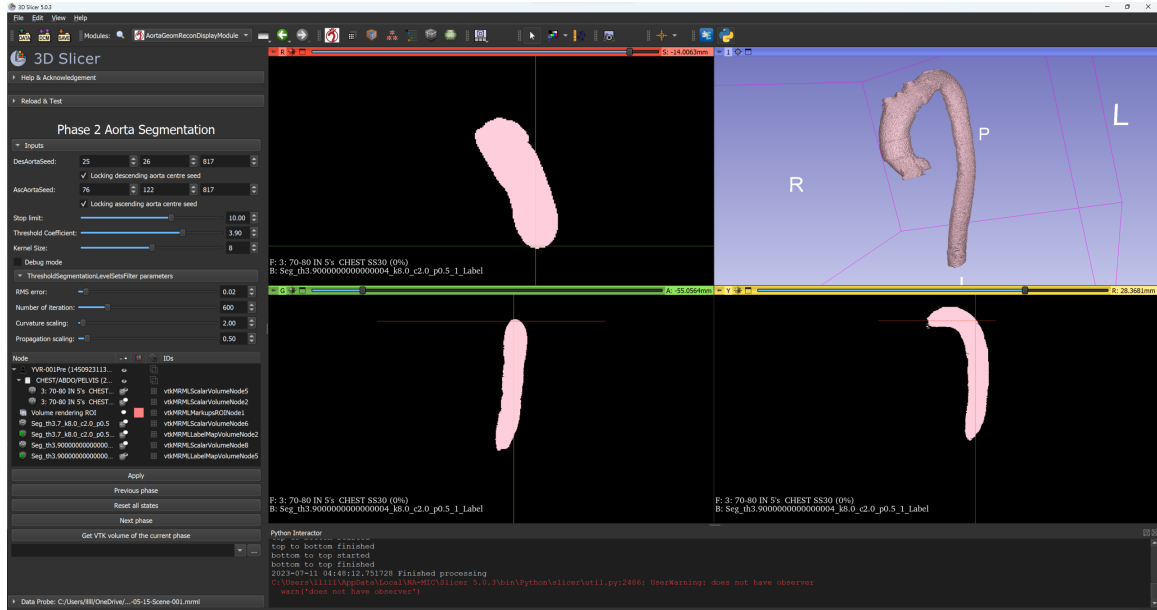


Figure 2.12: AortaGeomRecon UI phase 2

are stored as string. Every data stored in 3D Slicer can be accessed by the 3D Slicer's Widget Class and Logic Class for further processing.

On a higher level, 3D Slicer stores all the above data in a scene object. 3D Slicer can load any MRMLscene file, this allowed user to retrieve all the data nodes and parameters. On the other hand, 3D Slicer has a special input module, the DICOM database allowed user to store DICOM metadata in 3D Slicer.

2.3.2 3D Slicer's scripted module

Every ScriptLoadableModule in 3D Slicer have a Widget Class and a Logic Class. The Widget Class is used to initialize the extension module's UI component, and the parameters tied to the UI component. The module's Logic Class is used to perform the processing of the data. In the Logic Class, we initialize an AortaGeomRecon Segmenter object with the attributes set to the parameters reading from UI component,

which are inputs by the user. After completing the segmentation with Segmenter object, we convert the SimpleITK image object to a volume node corresponding in 3D Slicer, which allow the user to visualize the segmentation result.

2.4 GitHub and Workflows

This project uses GitHub for version control.

GitHub issues tracker use to keep track the items to work on throughout the development of the project

GitHub Project is used for dividing a large issue into smaller tasks, expected date of the completion. This is useful for project management.

GitHub Workflow is a great tool for Continuous Integration tests.

We uses GitHub Workflow for Linter and Continuous Integration tests.

Chapter 3

Assurance Cases and Selected Evidence for AortaGeomRecon

3.1 Assurance Case Development

3.2 Assurance Case for Software Specification Requirements

- ([Lin, 2023](#))
- Goals
- Assumptions
- Data Definitions
- Instance Model

- Functional Requirements
- Non-Functional Requirements
- Traceability Matrix Between Different Sections
- Traceability Matrix Between Requirements and Other Sections

3.3 Assurance Case for Implementation

- Design Document
 - Sphinx - Python Documentation Generator
 - Comments in the source code
 - Algorithm Overview - Explanation and Program Workflow
 - Glossary
- Module Guide
 - Module Decomposition
 - Traceability Matrix between Modules and Requirements
 - Traceability Matrix between Modules and Source Code
- Test Case
 - GitHub Workflow
 - Continuous Integration tests
 - * build “Ground Truth Data”

- * Steps
- * Coverage

3.4 Algorithm Review

3.4.1 Algorithm Review with Kailin Chu

3.4.2 Algorithm Review with Dr. Dean Inglis

- Python Spyder IDE - Data Visualization and Debugging tool
-
-

Chapter 4

Conclusion and Future Works

4.1 Thesis Summary

4.2 Future Works

([Kurugol et al., 2012](#)) discussed about a segmentation workflow that required less hyperparameters.

Appendix A

Your Appendix

Your appendix goes here.

Appendix B

Long Tables

This appendix demonstrates the use of a long table that spans multiple pages.

Col A	Col B	Col C	Col D
A	B	C	D
A	B	C	D
A	B	C	D
A	B	C	D
A	B	C	D
A	B	C	D
A	B	C	D

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Col A	Col B	Col C	Col D
A	B	C	D
A	B	C	D
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A	B	C	D
A	B	C	D
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A	B	C	D
A	B	C	D
A	B	C	D
A	B	C	D
A	B	C	D
A	B	C	D
A	B	C	D

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- Jingyi Lin. 2023. System Requirements Specification. <https://github.com/smiths/aorta/blob/main/docs/SRS/SRS.pdf>.