# Automatic Segmentation of the Diaphragm within CT Scans using Python

| Supervisor: F. Vidal | Department of Computer Science and Electronic Engineering | Bangor University Matthew Jones

#### Introduction The aim of this project is to fully diaphragm the from segment **Computed Tomography (CT) scans.**

CT images have low contrast, causing the diaphragm image intensity distribution to be similar to surrounding organs, such as liver, and spleen<sup>[1]</sup>. the heart, Segmentation of these surrounding organs can help towards the end goal of separating the diaphragm.

### DICOM File Format

DICOM (Digital Imaging and Communications in Medicine) is the standard format for medical images and has metadata containing information regarding the patient and the study. Pixel intensities of DICOM images represent the absorption/attenuation coefficient of radiation within a tissue; these intensities are typically analysed as Hounsfield units  $(HU; Figure 1)^{[2]}$ .



Figure 1. Hounsfield Scale<sup>[3]</sup>

### Key Tools

Python language was used to implement the SimpleITK API for reading DICOM files and applying different methods of segmentation to extract features from the data.





Thresholding, connected component image transforms filters, and neighbourhood operation techniques are used for initial segmentation and extraction of the lungs, ribcage (Figure 2, 3) and heart. Diaphragm contact points can then be found, and a B-Spline approximation technique applied to fully extract the diaphragm, as described within Figure 4.

Extraction of the heart and locating diaphragm contact points are the more challenging stages of the approach. Neighbourhood operations were used to label the heart. Diaphragm contact points relied upon applying various techniques for each feature (i.e. canny edge detection).



Figure 4. Key stages toward automatic diaphragm extraction from within CT scans and associated approaches.

#### Approach

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	Threshold (<120 HU)
•	Connected Component Filter
	Binary Threshold (>-950, <-300 HU)
•	Connected Component Filter
	Threshold within Lung sub- image (<35 HU)
	Neighbourhood Operation Identifying clumps
•	Connected Component Filter
•	Hough Transform
	Lungs: Locate arc-shaped curves at bottom surface of lungs
•	Heart: Canny Edge Detection
	Ribcage & Aorta: Morphological Image Closing; Neighbourhood operation
•	Multi-level B-Spline approximation



Figure 2. Original CT scan slices (Row 1). Ribcage segmentation (Row 2). Lung segmentation (Row 3).

#### Results & Future Research **Key Outputs:**

- 2D masks & 3D rendering of diaphragms output from approach (Figure 5)
- Stand-alone Python tool for automatically extracting the diaphragm from CT scans

#### **Measuring Success**

- Questionnaire & interview from expert
- Comparison against manual annotations

#### **Future Research**

Outputs from the work are expected to be developing relevant to experts biomechanical models and understanding of the diaphragm's physiology.

## References

[1] Karami E, Wang Y, Gaede S, Lee TY, Samani A. Anatomy-based algorithm for automatic segmentation of human diaphragm in noncontrast computed tomography images.Med Imaging (Bellingham). 2016;3(4):046004. doi:10.1117/1.JMI.3.4.046004

[2] DenOtter TD, Schubert J. Hounsfield Unit. In:StatPearls. Treasure Island (FL): StatPearls Publishing; May 11, 2020. [3] Elsayed, Omnia & Mahar, Khaled & Kholief, M. & Khater, Hatem. (2015). Automatic detection of the pulmonary nodules from CT images. 742-746. 10.1109/IntelliSys.2015.7361223.





Figure 3. 3D rendering of segmented ribcage & spine (Green) and lungs (Blue).



Figure 5. Envisaged final diaphragm segmentation output in 2D CT scan slices (a, b) and rendered in 3D (c, d) <sup>[3]</sup>.