

ABSTRACT

Image segmentation is a good way to analyze information in various fields of life. Image processing, especially image segmentation is very important and beneficial especially for the medical image segmentation and many other fields, from the application of segmentation know how the segmentation is important in our life. Image segmentation is the process of partitioning a digital image into sets of pixels. The aim of recognition system is to automatically identify the brain and extract the tumor from it. Several approaches have been proposed for medical segmentation. Some of the methods use color and brightness to reduce the complexity of the problem. Although such approaches solve the detect edges of regions. They are not able to handle almost any variation on the brain physical structure. There are many techniques have been proposed for tumor brain segmentation and Hill climbing is one of these techniques. The combination of the different approaches for the segmentation of brain images is presented in this project. Propose a color-based segmentation method that uses the K-means clustering technique with Hill climbing method to track tumor objects in magnetic resonance (MR) brain images, K-means clustering is used to cluster the image from gray to RGB scale while Hill climbing has been applied for the segmentation after that to overcome the problem with empty holes and Incoherent borders in the image , this project can successfully achieve segmentation for MR brain images to help pathologists distinguish exactly lesion size and region.

ABSTRAK

Segmentasi gambar adalah cara yang terbaik untuk menganalisis maklumat dalam pelbagai aspek kehidupan, terutama sekali dalam bidang perubatan. Segmentasi gambar adalah salah satu proses untuk mengasingkan bahagian-bahagian gambar digital ke dalam bentuk set piksel. Tujuan utama sistem pengenalan adalah untuk mengenal pasti otak dan juga ekstrak ketumbuhan. Beberapa pendekatan telah dikemukakan untuk bidang segmentasi perubatan. Terdapat beberapa kaedah yang menggunakan warna dan kecerahan untuk mengurangkan masalah. Walaupun terdapat pendekatan yang dapat mengesan masalah ruang, namun ia masih tidak mampu merangkumi hampir kesemua variasi dalam struktur fizikal otak. Terdapat pelbagai teknik yang telah dicadangkan di dalam kertas projek ini, antaranya ialah kaedah segmentasi warna yang menggunakan teknik k-means dan “hill climbing” untuk mengesan objek ketumbuhan pada magnet resonansi (MR). Kelompok k-means digunakan untuk mengklusterkan gambar daripada warna kelabu kepada skala RGB. Manakala kaedah “hill climbing” telah dilaksanakan dalam proses segmentasi dan seterusnya dalam menangani masalah tentang isu lubang-lubang kosong dan kecelaruan di dalam gambar. Projek ini berjaya dalam memperoleh segmentasi untuk mendapatkan gambar MR otak dalam membantu patologi membezakan saiz dan ruang yang tepat.

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CHAPTER 1

INTRODUCTION

1.1 Introduction

Segmentation is one of the first steps in an image analysis process, it is necessary and important. The main goal for the segmentation is to portioning the image in homogeneous regions in order to facilitate the scene interpretation which is done afterwards. That is why many segmentation methods have been proposed for the last years.

Image segmentation is one of most important task in image processing .It is used to analyze images in different fields; such as medical, science, agriculture and industry fields.

It is a great desire to have a fully automatic operation on image segmentation by the computer, because this process will speed up the process of image

segmentation. To date the manual segmentation is still widely used, due to a lot of constraints, these methods working depend on the place and the environment.

1.2 Problem Background

The segmentation methods are used to analyze images for specific purpose such as the medical and technology. Figure 1.1 illustrates the segmentation process on the human brain; the human brain has many parts, frontal lobe, olfactory bulb, temporal lobe, pons and cerebellum, as illustrated in figure 1.1, each one of these parts responsible to do special job.

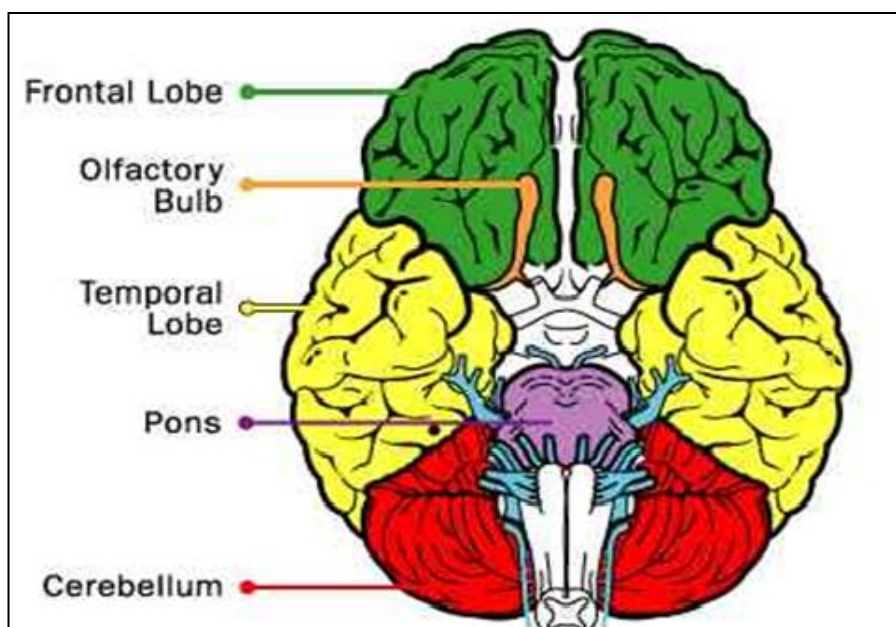


Figure 1.1 Illustrations of the human brain parts

The brain contains different parts with different shapes. The brain segmentation methods capable to recognize the tumor among all the brain parts, as illustrated in figure 1.2.



Figure 1.2 Segments the tumor in the brain

A recognition system aims to automatically identify the brain and extract the tumor from it. Several approaches have been proposed for medical segmentation. Some of the methods use color and brightness to reduce the complexity of the problem. Although such approaches solve the detect edges of regions. They are not able to handle almost any variation on the brain physical structure.

Segments the tumor from the brain is an important for to visualization the situation before do the surgery to achieve the desire result, it is important to raise the medical field to achieve the best result by using all the new techniques and by

utilization all the computer features to enhance the segmentation purpose. Also it is used the computer to speed up the procedure got doing the segmentation.

1.3 Problem Statement

Since the segmentation can be done by depending on the MRI, X-ray and many different ways, as shown in figures 1.3 and 1.4, these different methods have different levels of quality.

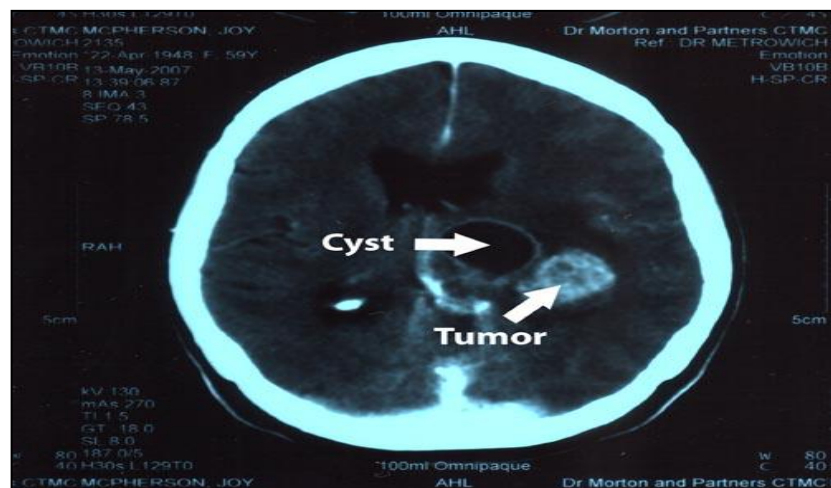


Figure 1.3 MRI for the tumor of the human brain

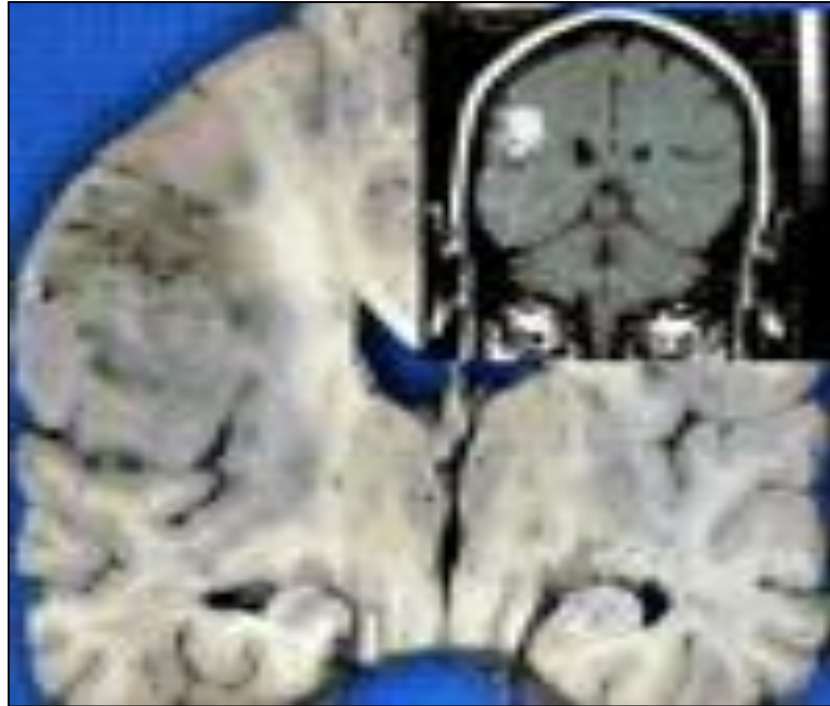


Figure 1.4 X-ray for the tumor of the human brain

The quality results depend on number of factors:

- Different methods.
- Problems with image acquisition.
- Poor quality of the MRI data.
- The homogeneous parts of brain.

The procedure can fail at any stage if the tumor is very homogeneous to the other brain parts. The human brain is greatly in size, shape, location, tissue composition and homogeneity has to be accounted, and because of these features, it's not easy to segment and to recognize it. The problem in this project is about distinguishing between the tumor and the other brain parts. To segment the tumor and ignore other parts.

1.4 Project aim

This project aims to design an enhancement of the Hill Climbing segmentation method to segment the tumor brain by using the MRI database, as shown in figures 1.5; 1.6. The approach is focusing on the tumor affected area only.

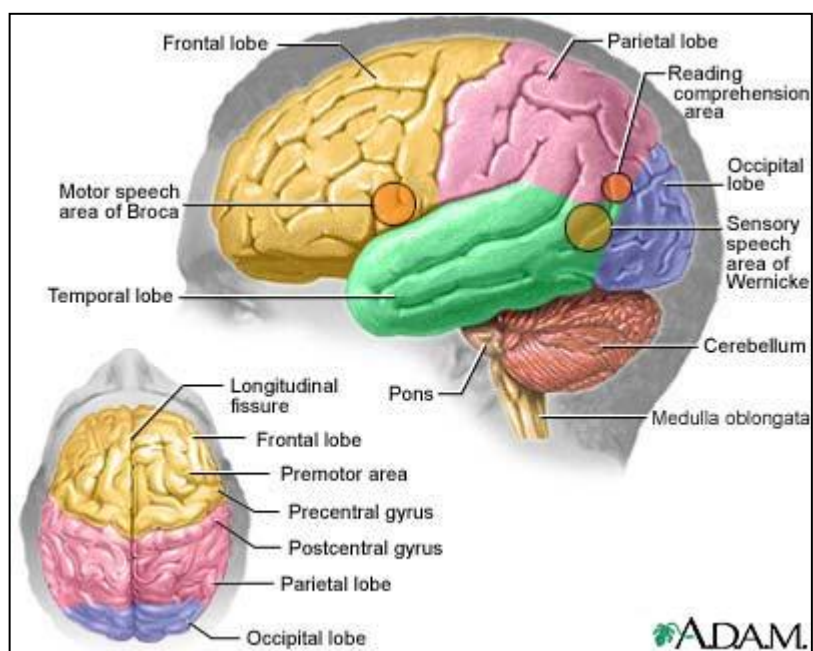


Figure 1.5 Human brain parts

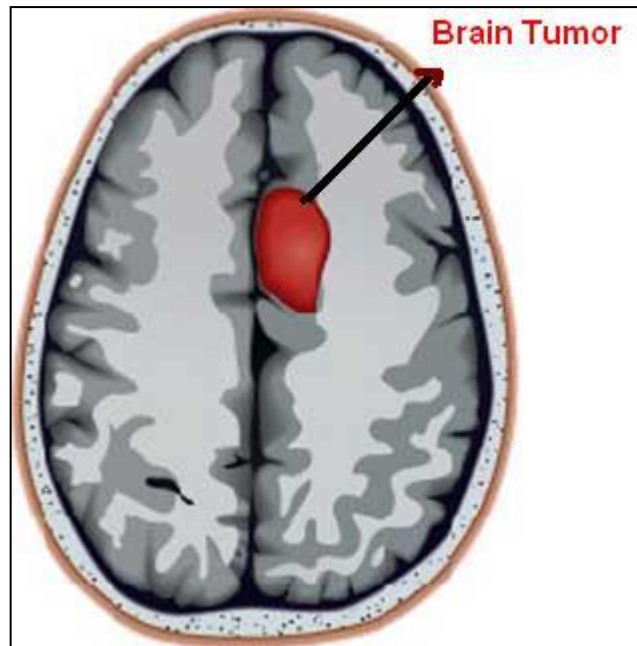


Figure 1.6 Segments the brain tumor

1.5 Objectives of the Project

This project intends to achieve the following objectives:

1. Study the previous methods of medical image segmentation.
2. To enhance the Hill climbing algorithm to segment the tumor.
3. To segment the brain tumor from the MRI data.

1.6 Project scopes

In order to achieve the objectives of the project, it is important to identify the scopes, which cover the following aspects:

- 1- The project will cover the methods and algorithms that solve the problem of tumor brain segmentation methods.
- 2- Find a technique by using an image in magnetic resonances so it will be capable to produce a complete division of image in separated region.
- 3- The project will not cover the other human brain parts.
- 4- Enhance existing algorithm of Hill Climbing method to achieve the objectives.
- 5- Brain images are taken from MRI database, the input data are from (Whole Brain Atlas) website for (Harvard University).

1.7 Project Organization:

The report is divided into 5 chapters:

- Chapter 1 describes the introduction and background of the study, as well as the project objectives, scope and the layout of this report.
- Chapter 2 gives literature reviews on the previous researches on the medical image segmentation, tumor features, the categories of image segmentation, some applications of segmentation, important factors in comparing segmentation methods, and the classification for the segmentation methods.
- Chapter 3 describes the project methodology that is been used.
- Chapter 4 gives the project results, result evaluation, overall discussion of the project and conclusion of the report.

- Chapter 5 gives the conclusion for the project and future works.

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