

SKIN TUMOR DETECTION THROUGH IMAGE PROCESSING

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ABSTRACT

In this paper a new method for skin tumor detection is developed using image processing. Considered a variation formulation, the energy of which combines a diffuse interface phase-field model (regularization term) and a log-likelihood computed using nonparametric estimates (data attachment term). Adopted CNN with the exact solutions which have the advantage to avoid space discretization and numerical instabilities. The resulting algorithm is simple and easy to implement in multi-dimensions. Concerning applications, focused on skin tumor segmentation. The clinical dataset used for the experiments is composed of 15 images with the ground truth given by a dermatologist. Comparisons with the reference methods, the proposed method is more robust to the choice of the volume initialization. Moreover, the flexibility introduced by the diffuse interface, the sensitivity increases by 12% if the initialization is inside the lesion, and the Dice index increases by 59%, if the initialization covers the entire lesion. The results show that this new method is well designed to tackle the problem of underestimation of tumor volumes.

INTRODUCTION

Digital Image is a non-destructive, fast, readily available and affordable means for determining the size, structure and echoic properties of skin lesions. This imaging modality is useful in the diagnosis, surgery planning and monitoring of benign normal malignant tumors of the skin, such as nevus, melanoma and basal cell. The vast majority of skin cancers are basal cell carcinomas and squamous cell carcinomas. While malignant, these are unlikely to spread to other parts of the body if treated early. They may be locally disfiguring if not treated early. However, it is well known that images suffer from two major defects, the low contrast and the presence of speckle noise making processing's more difficult. This explains why many works are still devoted to improve image segmentation algorithms. In this paper, we focus on the problem of skin tumor segmentation in digital images. Skin can also produce many types of cancers. Image processing is used to detect these tumors by using various methods like segmentation, filtering, feature extraction etc. To get an improved image or to get meaningful information from an image, it is necessary to convert an image into digital form and then perform functions on that image. It is a part of signal processing. The input is an image and it may be a video, a photograph and output is also another image having same characteristics as input image. Mostly Image Processing models take input samples as 2-D signals and after that they apply fixed signal processing methods to them. It is widely used technology now a days and it has various applications in the area of business.

LITERATURE SURVEY

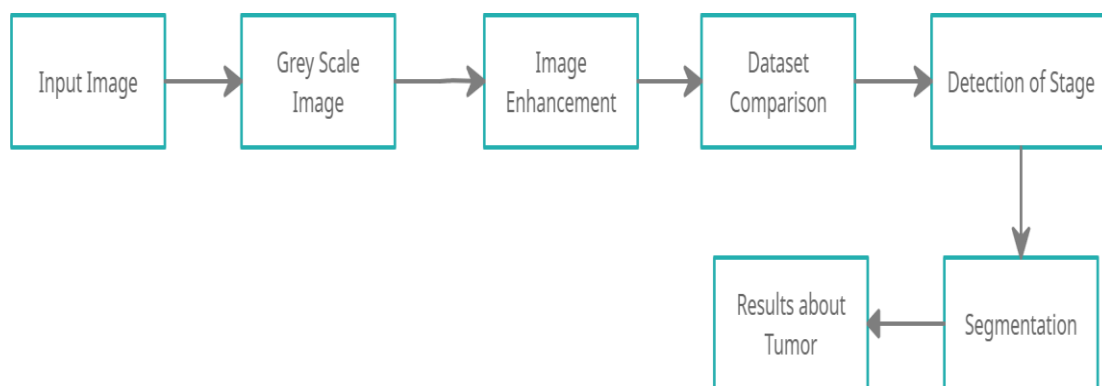
The skin cancer rate is increasing day by day because of pollution and climatic changes occurring in our environment. Skin tumors are frequent tumors to every person and various types of infections are becoming very frequent. You know that all of these tumors are very harmful, especially if not controlled at an early stage. Skin tumors can easily transfer from human to human so there is a need to control it their initial stage to prevent it from spreading. This skin cancer is basically of two major types they are melanoma and non-melanoma. This is mainly caused due to direct exposure to ultraviolet (UV) radiation due to pollution on ozone layer is depleting day by day which protects us from UV rays and now humans are directly exposed to ultraviolet rays based on a report every 10 percent depletion of the ozone layer will generate 3 million new non-melanoma and 4500 new melanoma cases. As per cancer.net statistics this year 1,00,350 people are diagnosed with melanoma type skin cancer in the United States out of this 60,190 are men and 40,160 are women. This skin cancer is ranked 5th most occurring type of cancer among men and the 6th most occurring type of cancer among women. This is more common in white people than black people the

average age of the infected people with this disease is 65. below 50 years fewer men are diagnosed than women with melanoma but if we consider the people with age above 65 the infected men are more than two times of infected women and for above 80 it is about three times, so by this, we can say that melanoma is more prevalent among the old age people but it is developed in young age people too. It is a common type of cancer diagnosed in young age adults in the group of young adults it is more in women than men this year in the United States 2,400 melanoma cases are diagnosed among the people between ages 15 to 19. This project presents an implementation of a skin tumors diagnosis system which helps user to detect human skin tumors and provides medical treatment timely. For this purpose, user will have to upload a tumor affected skin image to our system and give answers to the questions which are asked to user according to the symptoms of the skin. These symptoms are used to identify the tumor and provide a medical treatment. Non-Melanoma skin cancer is more common and mostly curable it is approximated that more than 3 million cases of non-melanoma are reported every year in the United States this number is rising every year. The rate of deaths is declining every year due to early detection and advance medical facilities. Most common types of non-melanoma are Basal cell carcinoma, Squamous cell carcinoma, Merkel cell. Every year Basal cell carcinoma is more diagnosed than the other two it constitutes more than 80% of non-melanoma skin cancer. Nearly 2000 people die every year with Basal carcinoma and Squamous cell carcinoma. The image of skin tumor is taken and various pre-processing techniques are applied onto that image for noise removal and image enhancement. This image is segmented by using a segmentation technique i.e. thresholding segmentation. At last, data mining techniques are used to identify the skin tumor and to provide recommendation to users.

1. Proposed Algorithm

Convolutional neural networks, also called ConvNets, were first introduced in the 1980s by Yann LeCun, a postdoctoral computer science researcher. LeCun had built on the work done by Kunihiko Fukushima, a Japanese scientist who, a few years earlier, had invented the neocognitron, a very basic image recognition neural network. The early version of CNNs, called LeNet (after LeCun), could recognize handwritten digits. CNNs found a niche market in banking and postal services and banking, where they read zip codes on envelopes and digits on checks. But despite their ingenuity, ConvNets remained on the sidelines of computer vision and artificial intelligence because they faced a serious problem: They could not scale. CNNs needed a lot of data and compute resources to work efficiently for large images. At the time, the technique was only applicable to images with low resolutions. In 2012, AlexNet showed that perhaps the time had come to revisit deep learning, the branch of AI that uses multi-layered neural networks. The availability of large sets of data, namely the ImageNet dataset with millions of labeled pictures, and vast compute resources enabled researchers to create complex CNNs that could perform computer vision tasks that were previously impossible.

2. Block Diagram

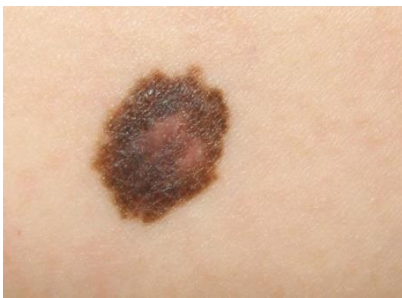


First, we will create a Graphical User Interface (GUI) to control the operation for the detection of tumor.

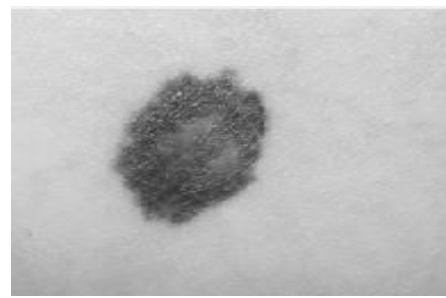
Here we will consider an input image where we have to detect the tumor. The image can be of any resolution, so we will change the resolution of the image into 256x256. The input image is in RGB image format so to extract the features from the tumor we have to convert the format of the image into Grey Scale image format. While capturing the image, there is a probability of occurrence of the noise in the image. So, to remove the noise from the image and to enhance the quality of the image we apply Discrete Wavelet Transform (DWT) method on the image. We will have a dataset which consists of few images and will apply the same procedure on each and every image in the dataset and we will train them for the detection of the tumor in the input image. Later we will compare the input image with the images in the dataset to detect the region of the tumor on the skin. Then we will segment the image in order to obtain the exact location of the tumor in the image. We will be able to calculate the area of the tumor in the image and also the type of tumor.

4.1 RGB to Grey-Scale Conversion

The input image is an RGB image. We are converting into Grey-Scale Image because color increases the complexity of the model. One may want to introduce an image processing tool using gray level images, as opposed to color, not because of the "format" of gray level images, but because their inherent complexity of gray level images is lower than that of color images.



InputImage



Grey-ScaleImage

4.2 Discrete Wavelet Transform Method

DWT is an algorithm used to reduce dimensionality of an image to reduce noise and to enhance picture quality during feature extraction process. In this project, we use Two Level Decomposition for analyzing 2-D signals like images. So 2-D input signal like digital images is projected on the four bases and the results are four different set of transform coefficients. 1. LL-Sub band consists of all wavelet coefficients, which is obtained from applying low pass filter to both rows and columns of an image. 2. HL-Sub band consists of all wavelet coefficients results from low pass filtering of the rows followed by high pass filtering of the columns. 3. LH-Sub band consists of all wavelet coefficients results from high pass filtering of the rows followed by low pass filtering of the columns and mostly it carries information about the vertical details or edges. HH-Sub band consists of all wavelet coefficients results from applying high pass filter to both rows and columns. This usually captures the diagonal edges or details of the original images.

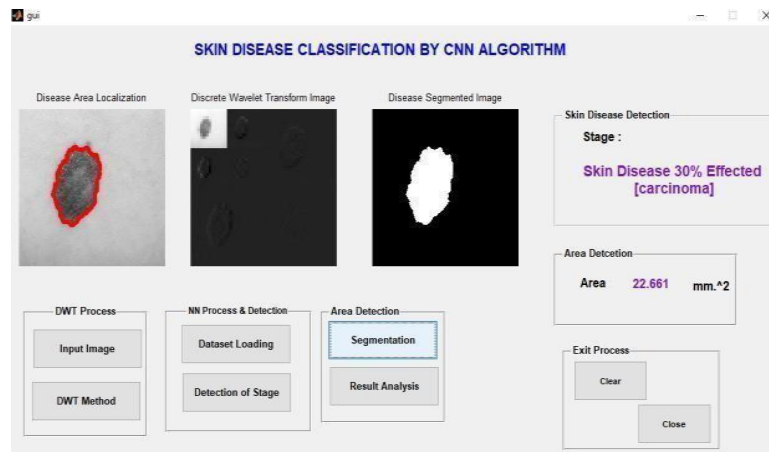


4.3 Image Segmentation

Image Segmentation is the process of partitioning a digital image into multiple segments. The goal of segmentation is to simplify and/or change the representation of an image into something that is more meaningful and easier to analyze. So with the help of Image Segmentation with will find the area of interest which is tumor region.



TumorDetection



3. Advantages

- The patient will get cure before it turns into cancer.
- We can provide awareness about the tumor and diagnosis for it.
- Low Cost.
- It can be used in Hospitals, Diagnostics Centers and Medical Research Centers.

4. Disadvantages

- This application is implemented only for three skin tumors (Basal Cell Carcinoma, Squamous Cell Carcinoma and Melanoma).
- It is implemented only for windows application so that is not yet developed for smart phones like Android, IOS etc.
- When capturing the image for this application, it is mandatory to capture it without any light effects.
- It only supports English language.

5. FUTURESCOPE

Due to the performance and less time complexity, proposed algorithms are useful in many applications in ultrasound skin imaging and Computer Aided Diagnosis system. In the further work, following issues can be investigated.

1. There are some images that the proposed algorithms cannot handle properly.
2. Proposed methods can be integrated with suitable classification algorithms to generate reliable Computer Aided Diagnosis system.
3. More US skin images can be examined to validate the performance of proposed methods.

8. CONCLUSION

In the proposed system, Image Pre-Processing, Image Segmentation for extracting the region of interest in the image itself and Image Classification steps are performed for categorizing skin lesion images into melanoma or carcinoma. We have detected the tumor using CNN algorithm where we will use DWT method (Two Level Decomposition) for removal of noise by not losing the data in the image. CNNs display a high performance as state-of-the-art skin lesion classifier. By using CNN algorithm, we are able to detect the tumor in the image with high accuracy and we are able to differentiate the tumor and the skin through these segmentation. The result encourages and motivates for future improvement and research for online diagnosing of melanoma in early stages before it is too late.

9. References

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