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LUNG FISSURE SEGMENTATION AND LOBE SEPARATION USING SVM CLASSIFICATION

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Abstract

Segmentation is chief method within the field of medical imaging, because it will give complete data of a picture. During this operation, segmentation of pulmonic lobe is allotted that is helpful for the clinical clarification of CT picture, to retrieve the first presence and also the characterization of many respiratory organ diseases. This segmentation method is exacting for very respiratory organ pathologic or respiratory organ with incomplete fissures. Existing strategies extremely rely on the detection of fissures whereas; this system becomes less reliable just in case of abnormalities. So as to cut back this, automatic segmentation of the respiratory organ lobe is completed victimization marker based mostly watershed rule and multi-atlas segmentation method. In a first step, inter lobular fissures are observed using a supervised enhancement filter. The fissures are then used to compute a cost image, which is incorporated in the watershed approach. By this, the segmentation is drawn to the fissures at places where structure data is present in the image. In areas with defective fissures (e.g. due to insufficient image quality or original conditions) the smoothing term of the level sets applies and a stopped continuation of the fissures is provided.

Key words: Segmentation, Pre-processing, Computed Tomography (CT), Fissure detection.

1. Introduction

The human lungs are divided into five distinct compartments called lobes. The separating junctions between the lobes are called the lobe fissures. The left lung consists of the upper and lower lobes, which are separated by the left oblique or major fissure. The right lung consists of the upper, middle, and lower lobes: the upper and middle lobes are separated by the horizontal or minor fissure; the middle and upper lobes and separated from the lower lobe by the right oblique which

is also known as major fissure as shown in (Fig-1). Relative lobar rotation of fissures to one another is allowed to adapt shape changes in the cavity. Each lobe is served by separate airway and vascular networks mostly in the case of incomplete fissures. In specific anatomic regions of the lung some pulmonary diseases are more common.

For example, tuberculosis and silicosis are fully upper lobe diseases, while interstitial pulmonary fibrosis is usually present in the lower lobes. Pulmonary emphysema is commonly present in the upper lobes, but in the case of lower lobes there is a rare genetic variant deficiency related with alpha-1 anti-trypsin. Thus, clinical important for disease classification and understanding the tissue and functional distinctiveness of the lobar parenchyma is more important which leads to the segmentation of lung lobes for effective identification of lung disease.

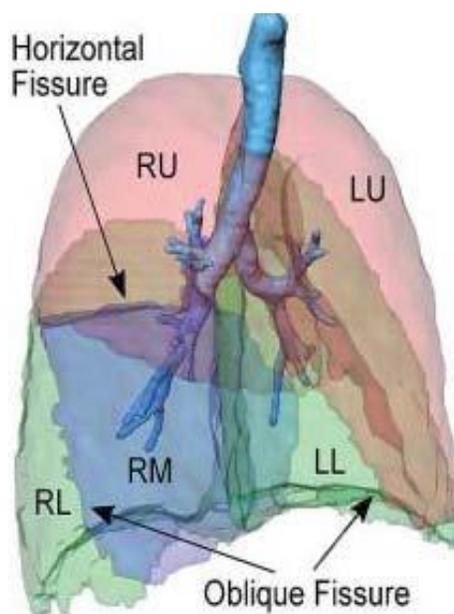


Figure-1.HumanLung.

Lobar analysis is vital in treatment choice, planning, and review as an example, there are rising therapies for respiratory illness that use full body part separation. These therapies think about computerized tomography (CT) pictures for characteristic the airway and body part allotment. Moreover, it should be necessary to spot incomplete fissures, since it's imagined to get collateral ventilation between lobes owing to incomplete fissures. Body part anatomy will be studied from CT imaging. Owing to low distinction and uneven form and look in CT imagination, it's a foremost challenge to the automated detection of the fissures that sometimes makes it tough even for manual analysts to mark their actual location. Usually, the fissures partition the lungs into 2 elements that seem as a skinny bright sheet. Some pulmonic diseases will modify the looks of the fissures on CT pictures that is given by Hayashi et al.

2. Input CT Scanned Image

A. Image Acquisition

First step is to amass the CT scan image of carcinoma patient. The respiratory organ CT pictures are having low noise compared to X-ray and imaging images; therefore they're thought-about for developing the technique. The most advantage of victimization X-raying pictures is that, it provides higher clarity and fewer distortion. For analysis work, the CT pictures are non-inheritable from NIH/NCI respiratory organ Image info association (LIDC) dataset. DICOM (Digital Imaging and Communications in Medicine) has become a regular for medical Imaging. Shows a typical CT image of carcinoma patient used for analysis. The non-inheritable pictures are in raw kind. Within the non-inheritable pictures ton of noise is observed to boost the distinction, clarity, separate the ground noise, it's needed to pre-process the photographs. Hence, numerous techniques like smoothing, improvement are applied to induce image in neededkind as shown in (Fig-2a).

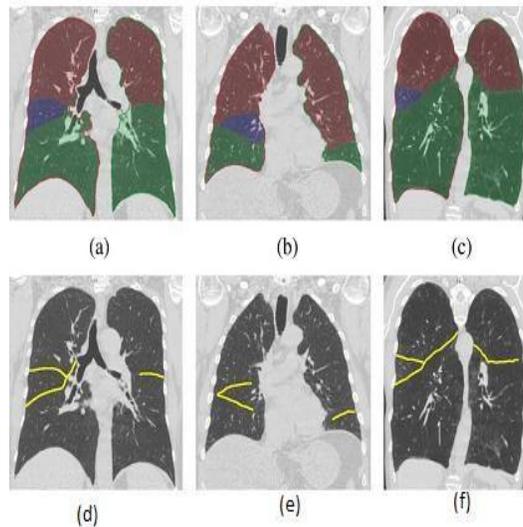


Figure-2. (a)Original image,(b)Blood vessel masking,(c)Level set approach,(d)Middle lobe segmentation, (e)Lower lobe segmentation,(f)Upper lobe segmentation.

Computed Tomography (CT) is taken into account in concert of the simplest strategies to diagnose the pneumonic nodules. It uses x-rays to get structural and practical info concerning the physical body. However, the CT image quality is influenced lots by the radiation dose. The standard of image will increase with the many quantity of radiation dose, however within the same time, this will increase the amount of x-rays being absorbed by the lungs. To forestall the physical body from all reasonably risk, radiologist's area unit obligated to cut back the radiation dose that affects the standard of image and is answerable for noises in respiratory organ CT pictures.

3. Pre-Processing

Pre-processing step aims to cut back the noises in these pictures. Totally different filtering techniques were planned in literature to get rid of these noises, like median filtering, wiener filtering Gaussian filter, bilateral filtering and a selected high-pass filter [11].

Several others works mix median filters with Laplacian filters by a differential technique, that subtracts a nodule suppressed image(through a median filter) from an indication increased image (through a Laplacian matched filter with a spherical profile)difference image, containing nodule increased signal, is then obtained and used for successive stages.

A.Masking

when running operations on the image the mask is used to restrict the result to the pixels that are 1 (selected, active, and white) in the mask. It is to simplify the representation of an image to easieranalyse as shown in (Fig-2b).

4. Lobe segmentation

The mean and maximum distances were calculated for each lobar border in 3-D by computing the distance between each voxel in the reference standard and the closest voxel in the lobar segmentation [12]. For cases with a poor lung segmentation, the volumetric overlay can be low even if the detection of the lobar borders completely correct as shown in(Fig-2d,e,f).

5. Watershed Transform

It is the tactic of alternative for image segmentation within the field of mathematical morphology. A grey-level image could also be seen as a geographic relief, wherever the grey level of a picture element is understood as its altitude within the relief.

A drop of water falling on a geographic relief flows on a path to finally reach a neighbourhood minimum. Naturally, the watershed of a relief corresponds to the bounds of the adjacent structure basins of the drops of water [12]. In image process, completely different watershed lines could also be computed. In (table-1,2) &(Fig-4,5)some could also be outlined on the nodes, on the perimeters, or hybrid lines on each nodes and edges. Watersheds can also be outlined within the continuous domain.

There also are many various algorithms to calculate watersheds. For a segmentation purpose, the length of the gradient is understood as elevation info.

Table-1: (Comparison ratio of Sensitivity and Mean Distance from center).

Sensitivity and Mean Distance from Center	Watershed Approach	Level Set Approach
Fissure Left	0.7	0.9
Fissure Right	0.85	1.24

Table-2: (Comparison ratio of Median Distance).

Median Distance	Watershed Approach	Level Set Approach
Fissure Left	9.97	11.2
Fissure Right	9.02	11.34

6. Input Image

The input pictures are unit chest CT scan pictures in JPEG format that contain tumors. 1st image designated from the file such by the string file name. The user needs to choose the specified respiratory organ CT scan image for any process. Then every image is re-sized to 256*256. Wiener filter in weight unit the input image is in RGB format. Thus its 1st reborn into grey scale image for any process. Then wiener filter of mask size 3*3 is employed to get rid of noise as a result of it's one in all the most effective strategies to get rid of the noise from the CT pictures; since these images sometimes contain artifacts or noise owing to patient movements.

Wiener filtering: The input image is in RGB format. In (Fig-14). Thus its 1st reborn into grey scale image for any process. Then wiener filter of mask size 3*3 is employed to get rid of noise as a result of it's one in all the most effective strategies to get rid of the noise from the CT pictures [13]; since these pictures sometimes contain artifacts or noise owing to patient movements.

7. Pre-Processing

A. Feature Extraction

In this method, total twelve textural options of all pictures within the info are extracted victimization GLCM (Gray level co-occurrence matrix). Then these options are used for tumor classification. GLCM is solely a matrix that offers the add

of the amount of times that the picture element with price i occurred within the particularized special relationship to a picture element with price j within the input image. Texture feature calculations use the contents of the [13] GLCM to allow a live of the variation in intensity at the picture element of interest.

B. Enhancement

Enhancement technique improves the contrast of images. The contrast enhancement can limit in order to avoid the noise which is present in an image as shown in (Fig-8).

8. Watershed Segmentation

In laptop vision, segmentation refers to the method of partitioning a digital image into multiple segments (sets of pixels, conjointly referred to as super pixels). Image segmentation is usually wont to find objects and limits (lines, curves, etc.) in pictures. additionally correctly, image segmentation is that the method of assignment a label to each constituent in a picture such pixels with constant label share sure visual characteristics.[8] The results of image segmentation could be a set of segments that conjointly cowl the complete image, or a collection of contours extracted from the image pixels in a very region is comparable with relevance some characteristic or computed property, like color, intensity, texture All image process operations usually aim at a higher recognition of objects of interest, i.e., at finding appropriate native options that may be distinguished from different objects and from the background. following step is to envision every individual constituent to examine whether or not it belongs to Associate in Nursing object of interest or not. This operation is termed segmentation and produces a binary image as shown in (Fig-10). A constituent has the worth one if it belongs to the article otherwise it's zero.

Image segmentation is that the action of partitioning a digital image into multiple segments. The goal of segmentation is to change or modification the illustration of an image into one factor that is plenty of purposeful and easier to research. Segmentation divides the image into its constituent regions or objects. The results of image segmentation could also be a collection of segments that along cowl the whole image or a set of contours extracted from the image [5]. Marker-controlled watershed segmentation supports this basic procedure and as shown in fig.3.2.3: 1) cipher a segmentation operate.

This may be an image whose dark regions ar the objects you make an endeavor to part. 2) cipher foreground markers. These ar connected blobs of pixels within each of the objects. 3) Cipher background markers. These ar pixels that are not

a locality of any object. 4) Modify the segmentation operate thus it alone has minima at the foreground and background marker locations. 5) Cipher the watershed rework of the modified segmentation operate.

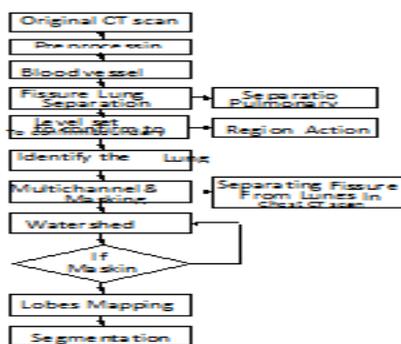


Figure-3. Flow diagram

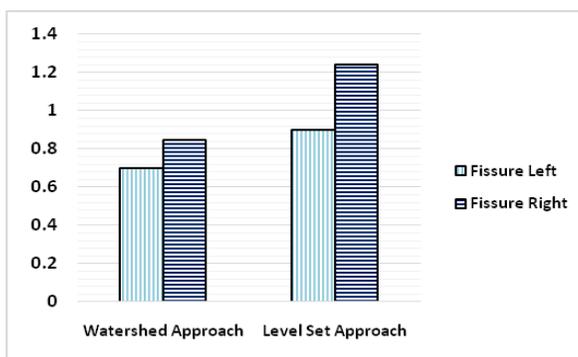


Figure-4. Comparison ratio of Sensitivity and Mean Distance.

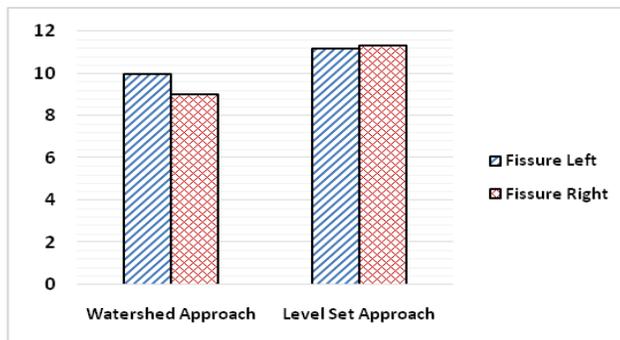


Figure-5. Comparison ratio of Median Distance.

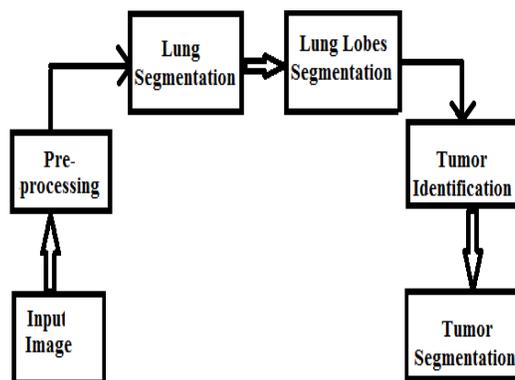


Figure-6. Block Diagram.



Figure-7. DICOM(Digital Image Communications in Medicine) image format.

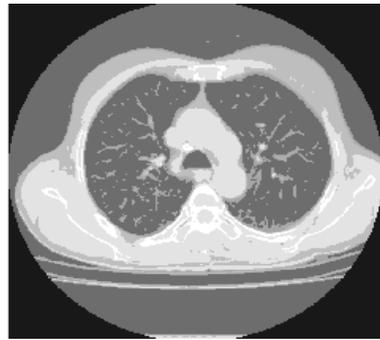


Figure-8.Enhanced image.

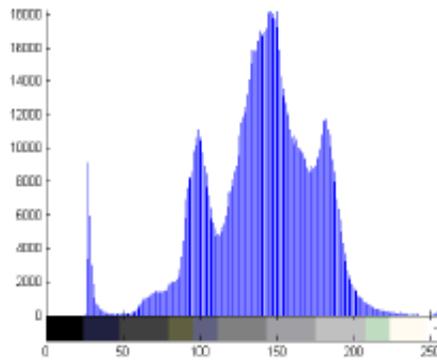


Figure-9.



Figure-10.Threshold binary image.

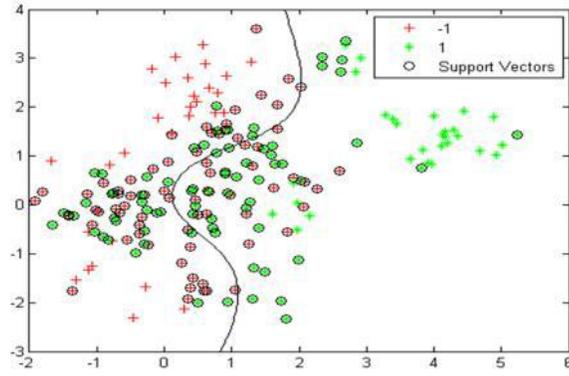


Figure-11. Support Vector Machine(SVM).

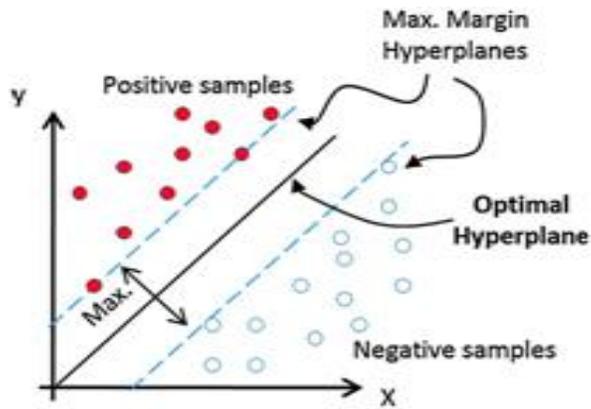


Figure-12. Maximum Margin Classifier.

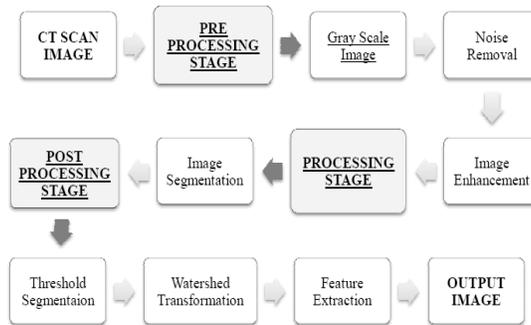


Figure-13. Flow Diagram.

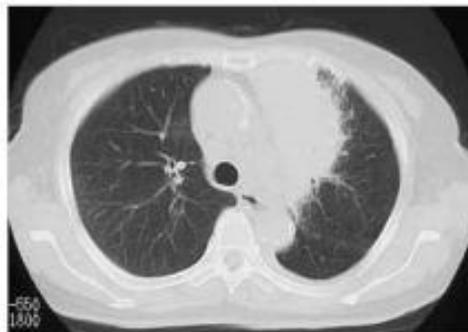


Figure-14. InputImage.

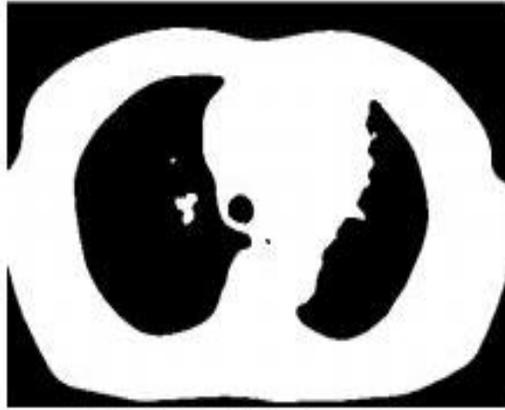


Figure-15.Output Image.

9. Histogram of Oriented Gradient (HOG)

HOG is a feature descriptor used in computer vision and image processing for the purpose of object detection.[13]The technique counts occurrences of gradient orientation in localized portions of an image. It is used for graphical representation of pixel intensity like angle, magnitude and gradient for a given image.

10. Support Vector Machine (SVM)

The Support Vector Machine (SVM) classifiers distinguish the non-tumorous (Kind) since tumorous (Hateful) lung knobs. It constructs a hyperplane in a great dimensional cosmos, which can be used for cataloging. The good parting is achieved by the hyperplane that has the major coldness to the nearest exercise facts point of any class. In general greater the margin lowers the simplification error of the classifier the story course is specified as determination to the classifier.[14] This technique distinguishes and identifies the non-tumorous (Benign) and cancerous (Malignant) lung distensions.

Technique and gratefulness randomly split database into 70% of the record for exercise and 30% for testing as shown in (Fig-11).Both subclass have the chance models from the same rotation. In implementation data, where both row agrees to an comment or imitate, and each column indicates to a feature or variable. The classifier sequence on the exercise set symbols it to the tiring set and then portion presentation by comparing the predicted markers and bounce conclusion as tumorous and non-tumorous.

A. Benign tumor

Benign tumors are non-tumors cells; but they need to be treated because they might harm the neighboring tissues or other vital organs.

B. Malignant tumor

Malignant tumors are cancerous cells and invade normal tissue or contain cancerous cells either from the lungs or other parts of the body.

11. Conclusion:

In this proposed system, a new method is used for the detection of fissures in the lungs from CT images using Watershed Segmentation, Histogram of oriented gradient (HOG), Support Vector Machine (SVM). And feature analysis, Enhancement and Feature Extraction are done in this method. Finally we can identify whether the fissures are normal or abnormal stage. If normal it is in benign stage of fissures. If it is in abnormal it is in malignant stage of fissures.

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