A Qualitative Review on Image Processing Algorithms to Detect Early Stage Lung Cancer

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Abstract: Nowadays, the image processing algorithms are being used widely in medical systems for detection of lung cancer. It is observed that the life span rate of lung cancer patients increases from 15 to 50% if they were detected at early stages. Detection of cancer cells is the most important issue for medical researchers as it becomes more complex in the treatment process. The detection steps of presence of cancerous cells include image pre-processing, segmentation, feature extraction and classification. In this paper, algorithms for enhancement, segmentation and feature extraction to detect the cancerous tumors which are small and large in size from the lung CT scan images are reviewed. Finally the algorithms are compared with one another using three parameters called accuracy, sensitivity and specificity.

Index Terms: CT Images, Image Preprocessing, Segmentation, Enhancement, Feature Extraction and Classification.

I. INTRODUCTION

In the human body the lungs exist as a pair of cone shaped organs on either side which look very smooth like a sponge. The right sided lung has three lobes, and is lightly larger than the left sided lung, which again has two lobes. The lungs take oxygen from the inhaled air and are transported to the bloodstream and to the other parts of the body using lung tissues and leaves the carbon dioxide when the air is exhaled. Lung cancer is a disease of abnormal blood cells which are multiplying and growing into a tumor within the lung. Cancer cells can be carried away from the lungs in blood, or lymph fluid that surrounds lung tissue. Metastasis occurs when a cancer cell leaves the site where it began and moves into a lymph node or to another part of the body through the bloodstream. Primary lung cancer is the cancer that just started in the lung.

Lung cancer symptoms consist of shortness of breath, wheezing, chest pain that does not get better, coughing accompanied with blood, difficulty in swallowing, and loss of weight and appetite. The most recent estimate statistics according to the American Cancer Society indicate that 226,160 new cases will be diagnosed (116,470 in men and 109,690 in women) in US, and there will be estimated 160,340 mortalities from lung cancer (87,750 in men and 72,590 among women). Furthermore, based on statistics from the World Health Organization (WHO), deaths caused by cancer will reach about 12 million people in 2030.

There are two main groups as small cell lung cancer and Non-small cell lung cancer

a) Small cell lung cancer: Recent study says that 20% of cancer cases come under this type. The primary cause for this type of cancer is Smoking. For this type doctors suggest the chemotherapy procedure instead of surgery.
b) Non-small cell lung cancer: This type of cancer is formed in the tissues of the lung. There are several types of non-small cell lung cancer. Signs of this type of cancer include a cough that arises very often and problem of effective breathing. The CT images are shown below for Lung with tumor in Fig.1

![Fig.1. CT Scan Image of Lung with a Nodule](image)

II. SURVEY OF RESEARCH WORK

The detection of cancerous tumor in the CT images includes four steps which is shown below in Fig.2.

![Fig.2. A Basic Lung Detection Process](image)

There are two image databases which are freely available for the researchers on medical image processing and they are Lung Image Database Consortium (LIDC) and ImageDatabase Resource Initiative (IDRI). At present, for the early detection of malignancy or cancer of lung and treatment in the early stage, the digital image processing approaches are most widely used and applied in medical fields in which the time factor is very important to discover the disease in the patient as soon as possible, especially in tumors. The Tumor stages is depicted in Table. I.

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Cancerous tumors are found in lung with the more people including men and women is quite common because of several reasons which include smoking, food habits, pollution and many more. There are certain methods available to detect cancerous cells at different stages of life or time. Diagnosis is mostly based on Computed Tomography (CT), MRI and PET images. The qualitative review presented in this paper focuses on detection of nodules and early symptoms of the diseases appearing in patient’s lungs. The first stage in detection is the preprocessing. The preprocessing of a CT image is done with help of the following filters:

1. Mean (or) average filter
2. Median filter
3. Adaptive median filter
4. Weiner filter

These filters are used to calculate the parameters like Mean Square Error (MSE), Peak Signal to Noise ratio (PSNR), etc. For segmentation process there are some articles are reviewed for evaluation tumor levels in the lung. For enhancement, and feature extraction there are certain approaches and algorithms are also reviewed in this paper. The required qualitative values are tabulated and analyzed and parameters are compared. The following three parameters are used in the analysis; they are accuracy, specificity and sensitivity which are tabulated from papers that are reviewed. The parameters are calculated with the following expressions.

\[
\text{Accuracy} = \frac{TP + TN}{TP + FP + TN + FN} \\
\text{Sensitivity} = \frac{TP}{TP + FN} \quad \text{and} \\
\text{Specificity} = \frac{TN}{TN + FP}.
\]

Where TP is True positive, FP is False Positive, TN is True Negative and FN is False Negative.

### III. RECENT ALGORITHMS PROPOSED

Image enhancement, segmentation, feature extraction and finally the classification methods that are reviewed is described below in brief for the tumors which are in big size.

Ada and Rajneet kaur [1] proposed a hybrid technique based on feature extraction and Principal Component Analysis (PCA) for lung detection in CT scan images. The image enhancement is implemented with histogram equalization and feature extraction and implemented by Binarization approach through which the features are extracted with number of white pixels having intensity more or less than the threshold value to check the normal or abnormal lungs. To predict the probability of lung cancer is done with binarization and GLCM approach.

Avinash et al. [2] proposed a method for detection of cancerous tumor in lung CT image using Gabor filter for Image enhancement. Instead of DWT and Histogram equalization, watershed segmentation algorithm was used and for feature extraction masking approach used instead of binarization. With this approaches they were able to improve the detection time of cancer.

Sanghamitra et al. [3] came with new method of detection of tumor with four stages of processing. First CT images are collected with different features, preprocessing is done with median filter, wiener filter or Gabor filter. Next, for segmentation, EK-means clustering method was used and the features extraction was done using gray-level co-occurrence matrix (GLCM) method and Back Propagation Network (BPN) was applied for classification of disease stages. They classified the given CT image as normal lung or cancerous lung. EK-Means clustering algorithm is one of the unsupervised clustering algorithm. In this method first clusters number ‘k’ is defined with an integer and randomly the cluster center is chosen. Then calculation of distance between each pixel and cluster center is done. The distance is specified as Euclidean function. Every single pixel is compared to all cluster centers using the distance formula. The pixel is moved to cluster which has shortest distance among all. Then the centroid is reevaluated. Again each pixel is compared to all centroids. The process continuous until the center converges. The performance rate of this algorithm depends on choice of Structural Element. Segmented lung image and its gray scale are shown in fig.3. K-Means clustering image is show in fig.4.
Md. Badrul et al. [4] gave another processing method in which the median filter was used to eliminate noise and to preserve the edges at the same time. They have proposed three types of threshold values. Threshold value ‘1’ is used to detect whether the lung is fully affected or not and ‘2’ and ‘3’ values are used to detect whether the left or right lungs are affected or not respectively. To calculate how accurate this technique compare to the others, they used Neural networks with some parameters as weight factors, and finally they concluded that the average accuracy was 96.67% with several tests based on the images collected from internet and hospitals as well.

Taruna et al. [5] presented a method for detection and classification of lung nodules from CT scan images. In this approach, a median filter is used for enhancing the image to get negligible noise and optimal thresholding with morphological closing operations. The gray level characteristics are used for finding the Region of Interest (ROI) i.e. lung tumor. Then from the segmented image geometrical features are analyzed by Linear Discriminative Analysis (LDA) classifier along with the statistical features. Finally, the results were given as accuracy of 84%, sensitivity of 97.4% and specificity of 53.3% which show that the prior classification of lung nodule can be detected effectively.

Elmar Rendon-Gonzalez and Volodymr Ponormaryov [6] defined the SVM (Support Vector Machine) Classifier based lung nodule detection system. The preprocessing involves selection of the threshold which is -500 HU followed by dilation and further erosion to get only lung image from the all other details. Using this approach, the False Positives can be reduced so that the sensitivity improves which further improves the accuracy of the tumor detection system. For the feature extraction GLCM matrices were used to give the gray level variations. The method estimates a function to classify the data in two classes. The decision function used was

\[
D(x) = W^T x + b
\]

where \(x\) is the training inputs, \(W^T\) is an \(m\)-dimensional vector, \(b\) is a bias term and for \(i = 1, ..., M\)

\[
w^T x_i + b \begin{cases} 
\geq 1 & \text{for } y_i = 1, \\
\leq -1 & \text{for } y_i = -1.
\end{cases}
\]

Support vectors are called the values which follow the above equations. This method provides the sensitivity of 84.93%, specificity of 80.92% and an accuracy performance of 78.08%. The performance of various methods for large tumors are depicted in Table II shown below.

**TABLE II. PERFORMANCE COMPARISON OF DETECTION ALGORITHMS (FOR THE TUMORS LARGE IN SIZE)**

<table>
<thead>
<tr>
<th>Method/Technique</th>
<th>Parameter</th>
<th>Accuracy</th>
<th>Sensitivity</th>
<th>Specificity</th>
</tr>
</thead>
<tbody>
<tr>
<td>Wavelet algorithm</td>
<td></td>
<td>79.45</td>
<td>87.67</td>
<td>83.56</td>
</tr>
<tr>
<td>Linear Discriminative Analysis (LDA)</td>
<td></td>
<td>84</td>
<td>97.14</td>
<td>53.3</td>
</tr>
</tbody>
</table>

For Image enhancement, segmentation, feature extraction and finally the classification methods described below in brief for the tumors small in size. Messay et al. [7] proposed an intensity thresholding with morphological processing for image segmentation and detection and for classification, they used Fisher Linear Discriminator (FLD) classifier for classification. With an average of 517.5 nodule candidates the method was able to detect 92.8% of all the nodules in the LIDC dataset. Overall, with a specificity of 3 false positives (FPs) per case/patient on average, the CAD system was able to correctly identify 80.4% of the nodules (115/143) using 40 selected features. A 7-fold cross-validation performance analysis using the LIDC database only shows CAD sensitivity of 82.66% with an average of 3 FPs per CT scan/case.

Cascio et al. [8] proposed a region growing algorithm followed by opening process for segmentation of selected image. The segmentation and the extraction of the suspected nodular lesions from CT images constitutes a new 3D segmentation, using deformable MSMS. For distinguishing the real nodules among nodule candidates, an additional classification step is applied; furthermore, a neural network is applied to reduce the false positives (FPs). The system performance was tested on a set of 84 images and detection rate with an accuracy of 97%. A reduction to 2.5 FPs/CT is achieved at 88% sensitivity.

Saleem Iqbal et al. [9] came with new idea in 4014 i.e. automatic detection and segmentation of lung small sized nodules, nodules in close vicinity of the diaphragm and lung wall and subsequent features extraction and classification for low and high contrast nodules on CT images. The methods include multistep threshold for the nodule detection and shape index threshold for false positive reduction. They used 60 CT scans of LIDC dataset and correctly detected 92% of total nodules in the case of 3-30 mm size nodules.

Mehdi Alilou et al. [10] used multiple thresholds followed by morphological opening and 3D region growing algorithm for segmentation of lung regions. Finally, a combination of a rule-based procedure and support vector machine classifier (SVM) were utilized to classify the nodules. The proposed method was applied on CT images of 60 patients, containing the total of 211 nodules, selected from the publicly available Lung Image Database Consortium (LIDC) image dataset. The proposed framework demonstrated acceptable detection performance of 80% (Sensitivity: 0.80; Fp/Scan: 3.9). After analyzing the recently proposed methods of detection of small size nodules a table is formed with the needed details.
The Gaussian filter for figure 5(b) is shown in figure 6. It determines the amount of smoothing, denoted as the standard deviation of the Gaussian distribution.

\[ G(x,y) = \frac{1}{(2\pi\sigma^2)} e^{-\frac{x^2+y^2}{2\sigma^2}} \]

where \(\sigma\) is the distance from the center on the horizontal axis, \(y\) is the distance from the center on the vertical axis, and \(\sigma\) is denoted as the standard deviation of the Gaussian distribution. It determines the amount of smoothing. The output of the Gaussian filter for figure 5(b) is shown in figure 6.

<table>
<thead>
<tr>
<th>Study (Year)</th>
<th>CT scan cases</th>
<th>Nodules</th>
<th>Nodules detected</th>
<th>Nodule size</th>
<th>Sensitivity (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cascio et al (2012)</td>
<td>84</td>
<td>148</td>
<td>130</td>
<td>&gt;3 mm</td>
<td>88</td>
</tr>
<tr>
<td>Saleem et al (2014)</td>
<td>60</td>
<td>222</td>
<td>211</td>
<td>3-30 mm</td>
<td>92</td>
</tr>
<tr>
<td>Alilou et al (2014)</td>
<td>60</td>
<td>211</td>
<td>169</td>
<td>&gt;4 mm</td>
<td>80</td>
</tr>
<tr>
<td>Gupta et al (2016)</td>
<td>60</td>
<td>315</td>
<td>301</td>
<td>1-30 mm</td>
<td>95.5</td>
</tr>
<tr>
<td>1010</td>
<td>3615</td>
<td>3058</td>
<td>1-30 mm</td>
<td>84.5</td>
<td></td>
</tr>
</tbody>
</table>

Fig 5. (a) Lung Without Tumor (b) Lung With Tumor

In the image pre-processing stage, image smoothing is the first step. For smoothing, Gaussian filter is applied on the input image. Gaussian smoothing is very effective for removing noise. Gaussian removes high frequency components from the image. So it is a low pass filter. Smoothing reduces the noise and giving us a more accurate intensity surface. The mathematical equation for the Gaussian filter is as given in equation 2.

Segmentation is used to divide an image into different small regions or objects. It has many applications in the medical field for the segmentation of the 2D medical images. After segmentation process, binarization process is done. In this approach, the total number of black pixels & white pixels are counted. If the total number of black pixels of input image is more than threshold, then the tumor is normal tumor. Otherwise, if the total number of black pixels is less than the threshold then the tumor is cancerous tumor.

Table III. Performance Comparision of Detection Algorithms (for the Tumors Small in Size)

Premlatha et al. [11] proposed a system consists of pre-processing, segmentation, feature extraction and final classification. The proposed marker controlled watershed segmentation technique separates the touching objects in the image. It provides best identification of the main edge of the image and also avoids over segmentation. It gives high percentage of accuracy compared to the thresholding algorithm. So it is efficient for segmentation. The proposed technique gives very promising results compared with other used techniques. The CT scan images which are used for processing are collected from the hospital. This image dataset contains lung CT scan images with tumor and without tumor. The Fig 5 shows some of the lung CT scan images with tumor and without tumor.

Fig 6. Output Image of Gabor Filter.

Fig 7. Marker-Controlled Watershed based Segmented Image

To improve the lung tumor detection quality certain methods were followed in the preprocessing. One such method is to use filters to make the image better and enhance it from noising, corruption or interference.

Vicky Ambule et al. discussed the about Median filters which can be used to reduce impulse noise level from corrupted images. Median filters are used to remove the salt-and-pepper noise. The median filter is a simpler nonlinear smoothing operation that takes a median value of the data inside a moving window of finite length. Median filter can be used to evaluate the averaging value of filter. A methodology based on median filter for the removal of salt and pepper noise by its detection followed by filtering in binary images has been proposed in this paper. An input image has been taken and tested with the approach of median filter. The CT image of tumor affected lung is shown in Fig. 8 which is further given to preprocessing using median filter. The image output of median filter is shown in the following Fig. 9.
case of Deep Neutral Network classification Algorithm and Sensitivity is high with the wavelet based Support Vector Machine and specificity is LDA classifier method in the case of large tumors. In the case of small size tumors a new approach having three masks in the segmentation followed by flood fill algorithm has given the average sensitivity is nearly equals to 90%. Hence there is a need of highly accurate algorithms at the segmentation, feature extraction and even in the last stage the classification stage.

REFERENCES


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IV. CONCLUSION

This paper focused on the recently algorithms developed by several authors along with the results which are tabulated above. Finally it is observed that the accuracy is more in the