

Detection of Lung Cancer Lesions Using 3D Convolutional Neural Networks and Segmentation for Accurate Detection

Rajani Kumari¹, C. Thanuja², K. Sai Thanvi³, K. Lakshmi ⁴, U. Lavanya ⁵

¹Assistant Professor, Department of Electronics and Communication Engineering, Sanskrithi School of Engineering Puttaparthi, Andhra Pradesh, India

^{2,3,4,5}B. Tech Students, Department of Electronics and Communication Engineering, Sanskrithi School of Engineering, Puttaparthi, Andhra Pradesh, India

ABSTRACT

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Accepted : 20 July 2021 Published : 27 July 2021 Lung cancer is a leading cause of death worldwide; it refers to the uncontrolled growth of abnormal cells in the lung. A computed tomography (CT) scan of the thorax is the most sensitive method for detecting cancerous lung nodules. A lung nodule is a round lesion which can be either non-cancerous or cancerous. In the CT, the lung cancer is observed as round white shadow nodules. In existing method, the candidate ROIs shape features are calculated, and some blood vessels are get rid of using rule-based according to shape features; secondly, the remainder candidates gray and texture features are calculated; finally, the shape, gray and texture features are taken as the inputs of the SVM (Support Vector Machine) classifier to classify the candidates. Experimental results show that the rule-based approach has no omission, but the misclassification probability is too large; Hence, in the proposed method the nodules were characterized by the computation of the texture features obtained from the gray level co-occurrence matrix (GLCM) in the wavelet domain and were classified using a SVM with radial basis function in order to classify CT images into two categories: with cancerous lung nodules and without lung nodules. The stages of the proposed methodology to design the CADx system are: 1) Extraction of the region of interest, 2) Wavelet transform, 3) Feature extraction, 4) Attribute and sub-band selection and 5) Classification. The same classification is implemented for the convolution neural networks. The final comparison is done between these two networks based on the accuracy.

Keywords : Lung Cancer, Convolutional Neural Network, Tensorflow, CT Scan

I. INTRODUCTION

Lung cancer is one of the most-fatal diseases all over the world today. About 1.8 million people have been suffering from lung cancer in the whole world [1]. In the United States, only 17% of people are diagnosed with lung cancer and they sur-vived for five years after the diagnosis. But the survival rate is lower in

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develop-ing countries [2]. The growth of uncontrolled cells can spread beyond the lung by the process of metastasis into nearby tissue or other parts of the body [3]. The cancer is localized to the lungs at the first two stages and is spread out to different organs in the latter stages. The diagnostic methods are CT scans (Computerized Tomography), chest radiography (Xray), MRI scan (Magnetic Resonance Im-aging) and biopsies etc. However, it is difficult to detect lung cancer in the early stage. In this research, we have collected CT scan images of 1500 patients. But we have worked on the CT images of 100 patients where each of them contains more than 120 DICOM 3D images. Among these, 80 patients' images are used here for training purposes and 20 patients' images are used for testing purpose.Our system is robust as well as effective for the early detection of lung cancer.

This research contributes to the following:

1) A literature survey is performed on the existing state-of-the-art techniques

for the detection of lung cancer.

2) A comprehensive study is performed with standard dataset using deep

convolutional neural network architectures for lung cancer detection in the early

stage.

II. LITERATURE REVIEW

S. Logesh kumar1, M. Swathy, S. Sathish, J. Sivaraman and M. Rajasekar, "Identification of Lung Cancer Cell using Watershed Segmentation on CT Images", Indian Journal of Science and Technology, Vol9(1),DOI:10.17485/ijst/2016/v9i1/85765, January 2016.

- In this experiment, various Computed Tomography (CT) images of the lung were used as input images with pixel size of 512×512 and obtained the output image in JPEG format.
- Compared to X-ray CT images are perceptive characteristics of identifying lung tumor size and lymph nodes.

- Thresholding selects a threshold value T and it assigns two levels to the image that is above value and below value for original threshold value.
- FFT, Auto enhancement and Gabor filtering are the three methods used as image enhancement techniques.
- Based on this study it is more evident that watershed segmentation technique is good for evaluation of lung cancer cell region.

III. Methodology

1) Convolutional layer:

This layer is where images are translated into featuremap data by convolu-tional kernels or filters. In a 3D CNN, the kernels move through three dimen-sions of data (height, length, and depth) and produce 3D maps. A 3D CNN is necessary for analyzing data where temporal or volumetric context is important.

2) Pooling layer:

Pooling, or down-sampling, is done on the convolutional output. During pooling, a filter moves across the convolutional output to take either the average or the weighted average or the maximum value. The goal of the pooling layer is to progressively reduce the spatial size of the matrix to reduce the number of pa-rameters and to control over fitting.

3) Fully-connected layer:

The fundamental goal of a fully connected layer is to take the results of the convolution and pooling processes and use them to classify the image into a label. In this layer, a softmax function is used to get probabilities as it pushes the values between 0 and 1. Batch normalization is used to improve the training speed and to reduce overfitting.

3.2. Proposed Methodology:



The proposed method consists of the convolution neural network which is a deep learning technique used for classification processes. The convolution neural network and its layers is explained below:

This architecture consists of 7 layers. The input images are fed to the network taking 50 images as a batch at a time. All images are resized to have 32 x 32 dimensions and all these images are true color or RGB images. And the layers of DCCN are three convolutional layers, three pooling layers and a fully connected layer.



Fig.3.1:Block Diagram of Proposed Method

3.3. Datasets:

We used LUNA16 (Lung Nodule Analysis) datasets (CT scans with labeled nodules). The LUNA 16 dataset has the location of the nodules in each CT scan. Thus, it will be useful for training the classifier. The inputs are the image files that are in "DICOM" format. The format and configuration of the images are different since the images are captured at different time and from different types of the camera. Actually, the images are of size $(z \times 512 \times 512)$, where z is the number of slices in the CT scan and varies depending on the resolution of the scanner [13]. Such large images cannot be fed directly into convolutional neural network architecture because of the limit on the computation power. At first, we converted all the images into similar size and format. To reduce the size of the input data, we have segmented the image. Thus, we have to find the regions that are more probable to have cancer. We have reduced our search space by first segmenting the lungs and then removing the low intensity regions. We have used the pixel as input to the neural network. The total size of the input data was 15 GB. A small subset of data of size around 2 GB has used for various testing

DWT band1 DWT band2
DWT band3 DWT band4
DWT band4

IV. Experimental Results





Fig.4.2:shows Tumor present around corners of lung



Fig.4.3:Input (CT image) and output images for lung where tumor is present in bronchioles

purposes.



V. Future Work

In this research, we used a vanilla 3D CNN classifier to determine whether a CT image of lung is cancerous or non-cancerous. Before using the 3D CNN, we preprocessed the CT image through a thresholding technique. We have performed a thorough experiment using LUNA 16 dataset. Our obtained detection accuracy is 80%, which is better than existing methods. In future, we will perform the experiments on a large amount of data and apply more features such as nodule size, texture and position for further improvement. We will also try to apply the state-of-the-art deep CNN methods for higher accuracy and use our method on other types of cancer detection.

VI. Conclusion

The 3D structure of nodules is used to extract features by 3D CNN to predict lung cancer. By using CNNs, the tedious task of manually extracting features can be eliminated. The obtained precision and recall obtained for the model are decent considering that less labelled data than most state-of-the-art CAD systems are used.

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