

Lung Cancer Detection Using Conventional Neural Network

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Abstract Lung most cancers is the main purpose for cancer-related death. Lung most cancers can provoke in the windpipe, major airway or lungs. It is brought about by way of unchecked increase and unfold of some cells from the lungs. People with lung disorder such as emphysema and preceding chest troubles have greater danger to be identified with lung cancer. Over utilization of tobacco, cigarettes and beedis, are the fundamental chance thing that leads to lung most cancers in Indian men; however, amongst Indian women, smoking is now not so common, which point out that there are different elements which lead to lung cancer. Other danger elements consist of publicity to radon gas, air-pollutions and chemical compounds in the workplace.

Lung most cancers detection at early stage has come to be very vital and additionally very handy with photograph processing and deep getting to know techniques. In this find out about lung affected person Computer Tomography (CT) scan pix are used to discover and classify the lung nodules and to notice the malignancy stage of that nodules. In this challenge we are the usage of CNN algorithm to become aware of Lung most cancers from CT-SCAN pics and to educate CNN we have CT-SCAN photographs dataset.

1.INTRODUCTION

It is most common in smokers accounting 85% of cases among all. So many Computer Aided Diagnosis (CAD) Systems are developed in recent years. Detection of lung cancer at early stage is necessary to prevent deaths and to increase survival rate. Lung nodules are the small masses of tissues which can be cancerous or noncancerous also called as malignant or benign. Benign tissues are most commonly non-cancerous and does not have much growth where malignant tissues grows very fast and can affect to the other body parts and are dangerous to health.

1.1 SCOPE:

For medical imaging so many different types of images are used but computer Tomography (CT) scans are generally preferred because of less noise. Deep learning is proven to be the best method for medical imaging, feature extraction and classification of objects. Several types of deep learning architectures are introduced by so many researchers to classify the lung cancer.

2.LITERATURE SUREVY

2.1 An Automatic Detection System of Lung Nodule Based on Multi-Group Patch-Based Deep Learning Network

AUTHORS: Hongyang Jiang, He Ma, Wei Qian, Mengdi Gao and Yan Li

ABSTRACT: High-efficiency lung nodule detection dramatically contributes to the risk assessment of lung cancer. It is a significant and challenging task to quickly locate the exact positions of lung nodules. Extensive work has been done by researchers around this domain for approximately two decades. However, previous computer-aided detection (CADe) schemes are mostly intricate and time-consuming since they may require more image processing modules, such as the computed tomography image transformation, the lung nodule segmentation, and the feature extraction, to construct a whole CADe system. It is difficult for these schemes to process and analyze enormous data when the medical images continue to increase. Besides, some state of the art deep learning schemes may be strict in the standard of database. This study proposes an effective lung nodule detection scheme based on multigroup patches cut out from the lung images, which are enhanced by the Frangi filter. Through combining two groups of images, a four-channel convolution neural networks model is designed to learn the knowledge of radiologists for detecting nodules of four levels. This CADe scheme can acquire the sensitivity of 80.06% with 4.7 false positives per scan and the sensitivity of 94% with 15.1 false positives per scan. The results demonstrate that the multigroup patch-based learning system is efficient to improve the performance of lung nodule detection and greatly reduce the false positives under a huge amount of image data.

2.2 Deep residual learning for image recognition

AUTHORS: K. He, X. Zhang, S. Ren, J. Sun

ABSTRACT: Deeper neural networks are more difficult to train. We present a residual learning framework to ease the training of networks that are substantially deeper than those used previously. We explicitly reformulate the layers as learning residual functions with reference to the layer inputs, instead of learning unreferenced functions. We provide comprehensive empirical evidence showing that these residual networks are easier to optimize, and can gain accuracy from considerably increased depth. On the ImageNet dataset we evaluate residual nets with a depth of up to 152 layers - 8× deeper than VGG nets [40] but still having lower complexity. An ensemble of these residual nets achieves 3.57% error on the ImageNet test set. This result won the 1st place on the ILSVRC 2015 classification task. We also present analysis on CIFAR-10 with 100 and 1000 layers. The depth of representations is of central importance for many visual recognition tasks. Solely due to our extremely deep representations, we obtain a 28% relative improvement on the COCO object detection dataset. Deep residual nets are foundations of our submissions to ILSVRC & COCO 2015 competitions¹, where we also won the 1st places on the tasks of ImageNet detection, ImageNet localization, COCO detection, and COCO segmentation.

2.3 Accurate Pulmonary Nodule Detection in computed Tomography Images Using Deep Convolutional Neural Networks

AUTHORS: Jia Ding, Aoxue Li, Zhiqiang Hu, Liwei Wang

ABSTRACT: Early detection of pulmonary cancer is the most promising way to enhance a patient's chance for survival. Accurate pulmonary nodule detection in computed tomography (CT) images is a crucial step in diagnosing pulmonary cancer. In this paper, inspired by the successful use of deep convolutional neural networks (DCNNs) in natural image recognition, we propose a novel pulmonary nodule detection approach based on DCNNs. We first introduce a deconvolutional structure to Faster Region-based Convolutional Neural Network (Faster R-CNN) for candidate detection on axial slices. Then, a three-dimensional DCNN is presented for the subsequent false positive reduction. Experimental results of the LUn

Nodule Analysis 2016 (LUNA16) Challenge demonstrate the superior detection performance of the proposed approach on nodule detection (average FROC-score of 0.893, ranking the 1st place over all submitted results), which outperforms the best result on the leaderboard of the LUNA16 Challenge (average FROC-score of 0.864).

4.PROPOSED WORK

Lung cancer identification at an early stage has become extremely crucial, as well as quite simple, thanks to image processing and deep learning techniques. Lung patient Computer Tomography (CT) scan images are used in this study to locate and classify lung nodules, as well as to determine the malignancy level of those nodules. In this research, we are utilising the CNN algorithm to detect lung cancer from CT-SCAN images, and we have a dataset of CT-SCAN images to train the CNN. The primary goal of this research is to investigate the performance of a classification algorithm in order to make an early diagnosis of lung cancer.

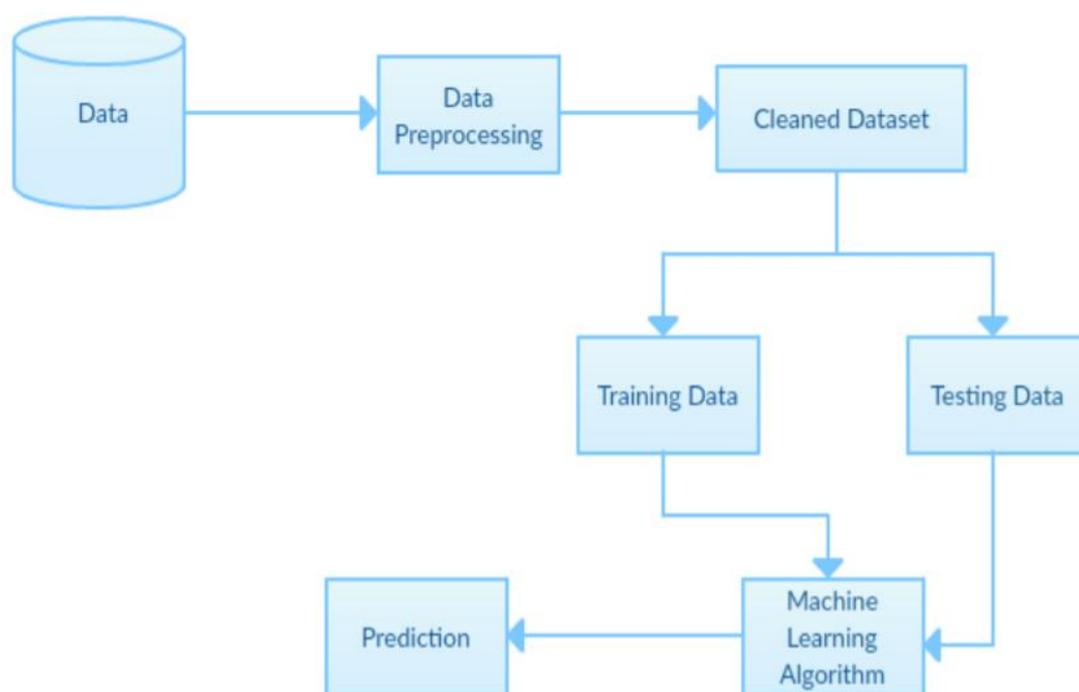


Fig 1: System Architecture

3.1 IMPLEMENTATION

- Upload Lung Cancer Dataset
In this module use upload dataset
- Preprocess Dataset
In this module data undergoes preprocessing
- Model Generation
In this module model generation is take place to predict disease.
- Build CNN Model

In this module CNN model is build.

- Accuracy & Loss Graph
In this module comparison graph is shown
- Upload Test Image & Predict Cancer

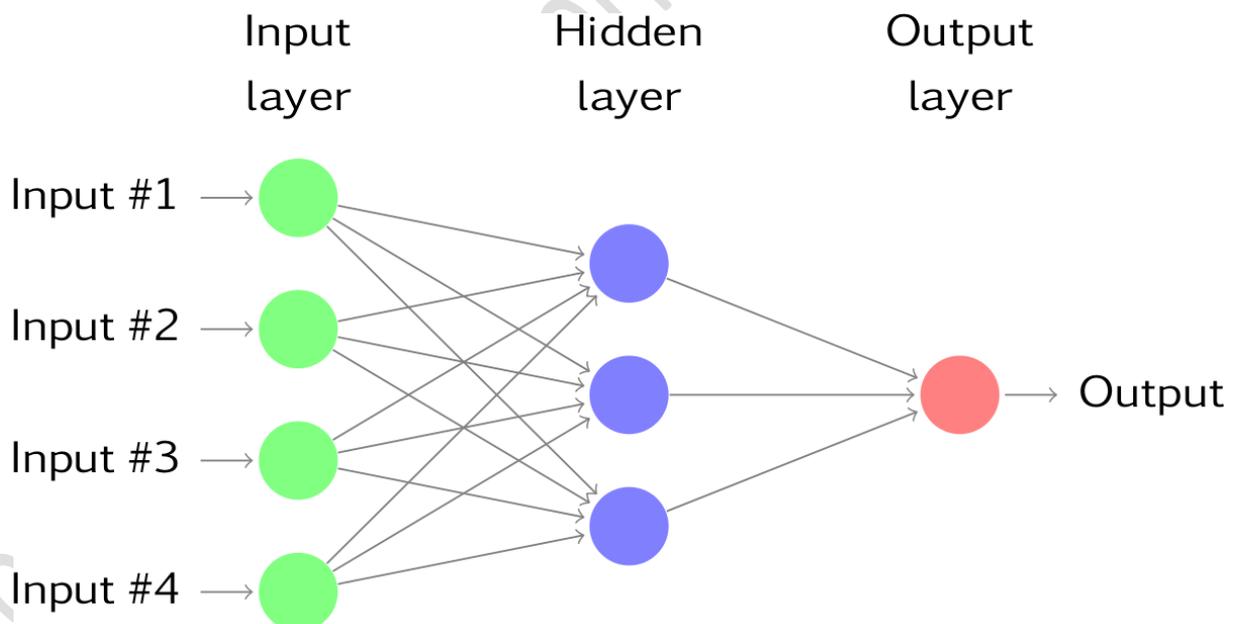
In this module, user uploads test image to predict diseases.

3.2 Convolutional neural network:

To demonstrate how to build a convolutional neural network based image classifier, we shall build a 6 layer neural network that will identify and separate one image from other. This network that we shall build is a very small network that we can run on a CPU as well. Traditional neural networks that are very good at doing image classification have many more parameters and take a lot of time if trained on normal CPU. However, our objective is to show how to build a real-world convolutional neural network using TENSORFLOW.

Neural Networks are essentially mathematical models to solve an optimization problem. They are made of neurons, the basic computation unit of neural networks. A neuron takes an input (say x), do some computation on it (say: multiply it with a variable w and adds another variable b) to produce a value (say; $z = wx + b$). This value is passed to a non-linear function called activation function (f) to produce the final output(activation) of a neuron. There are many kinds of activation functions. One of the popular activation function is Sigmoid. The neuron which uses sigmoid function as an activation function will be called sigmoid neuron. Depending on the activation functions, neurons are named and there are many kinds of them like RELU, TanH.

If you stack neurons in a single line, it's called a layer; which is the next building block of neural networks. See below image with layers



To predict image class multiple layers operate on each other to get best match layer and this process continues till no more improvement left.

Deep learning not only accelerates the critical task but also improves the precision of the computer and the performance of CT image detection and classification.

In this paper, the problem of classification of benign and malignant is considered. It is proposed to employ, respectively, the convolution neural network (CNN) and deep neural network (DNN).

The input data (image data) has a strong robustness on the distortion. The multiscale convolution image feature is generated by setting the convolution kernel size and parameter; the information of different angles is generated in the feature space.

4.RESULTS AND DISCUSSION

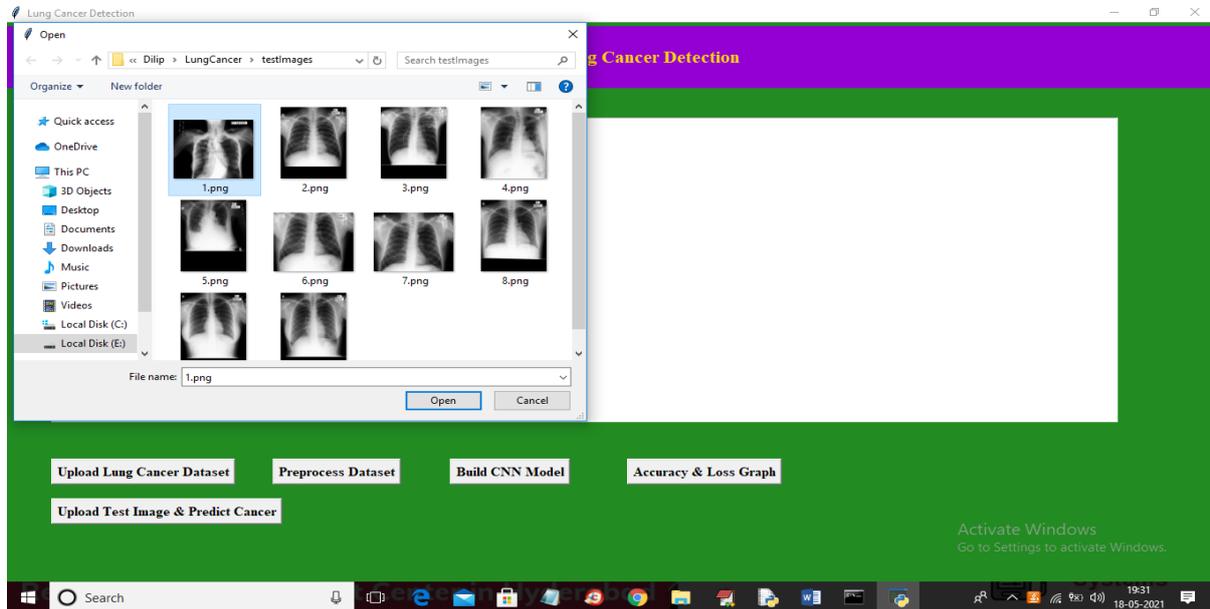


Fig 2:In above screen selecting and uploading ‘1.png’ file and then click on ‘Open’ button to get below result

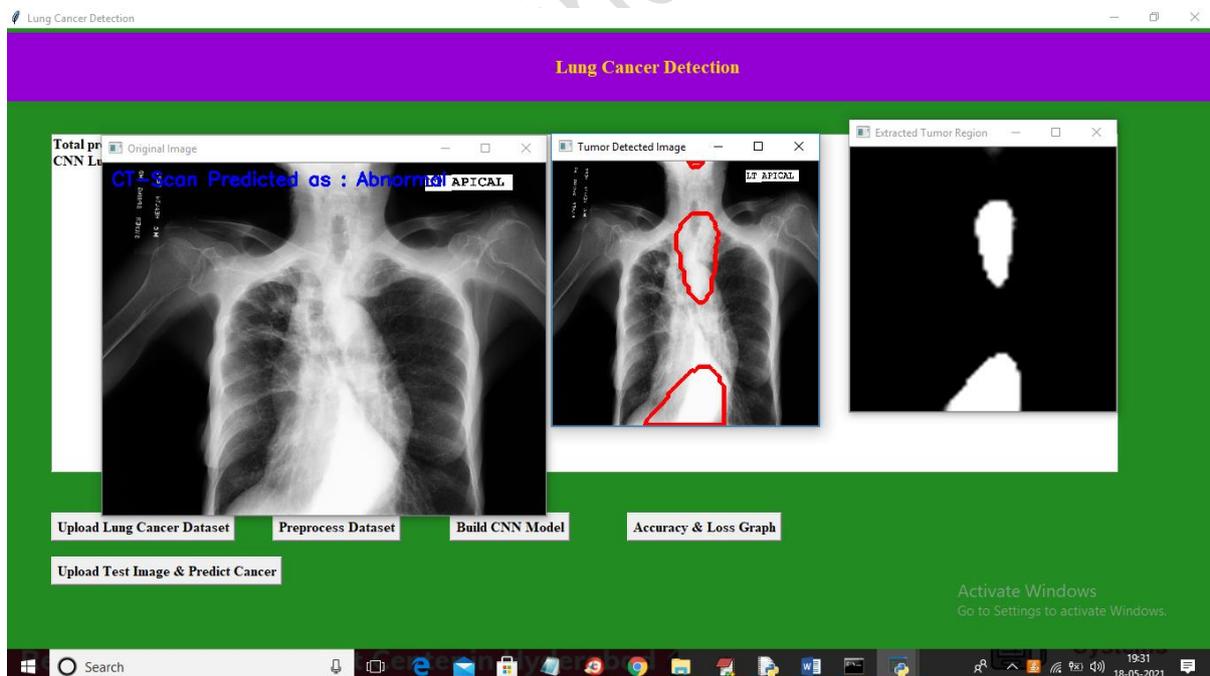


Fig 3:In above screen in first image in blue colour text we can see predicted result as CT-SCAN contains abnormality and in second image we are detecting places were abnormality detected and in third image we extracted all abnormality patches from original image and then displaying. Now test other image

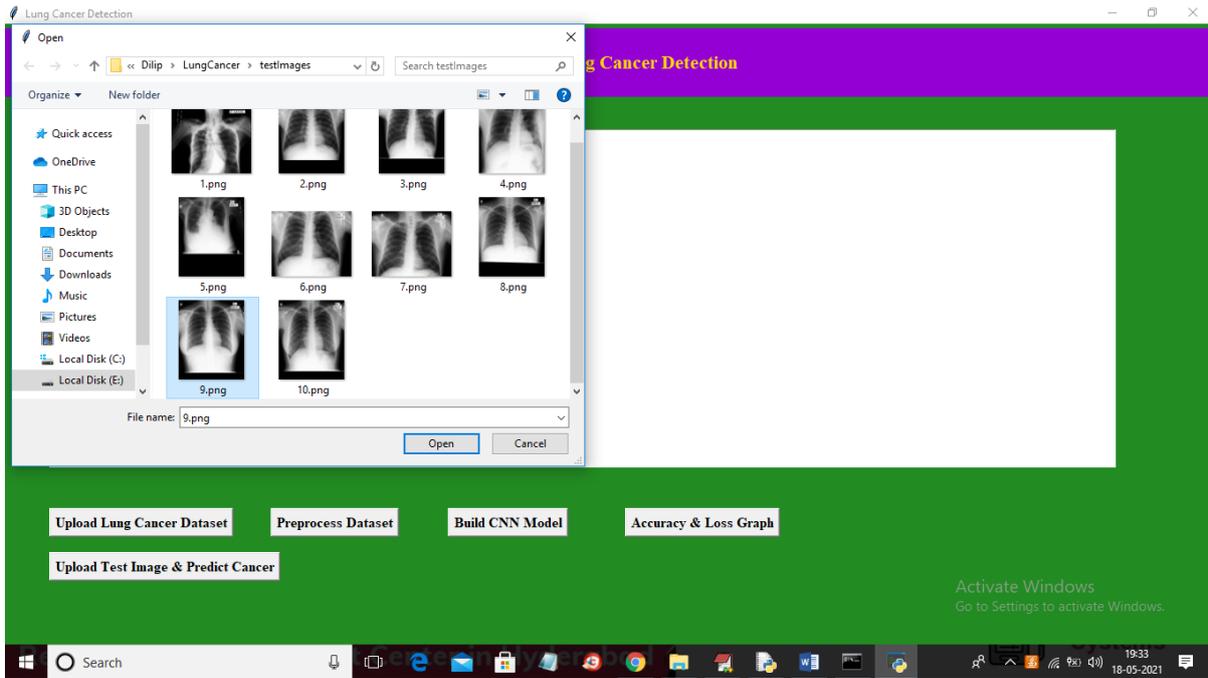


Fig 4:In above screen selecting and uploading '9.png' file and then click on 'Open' button to get below result

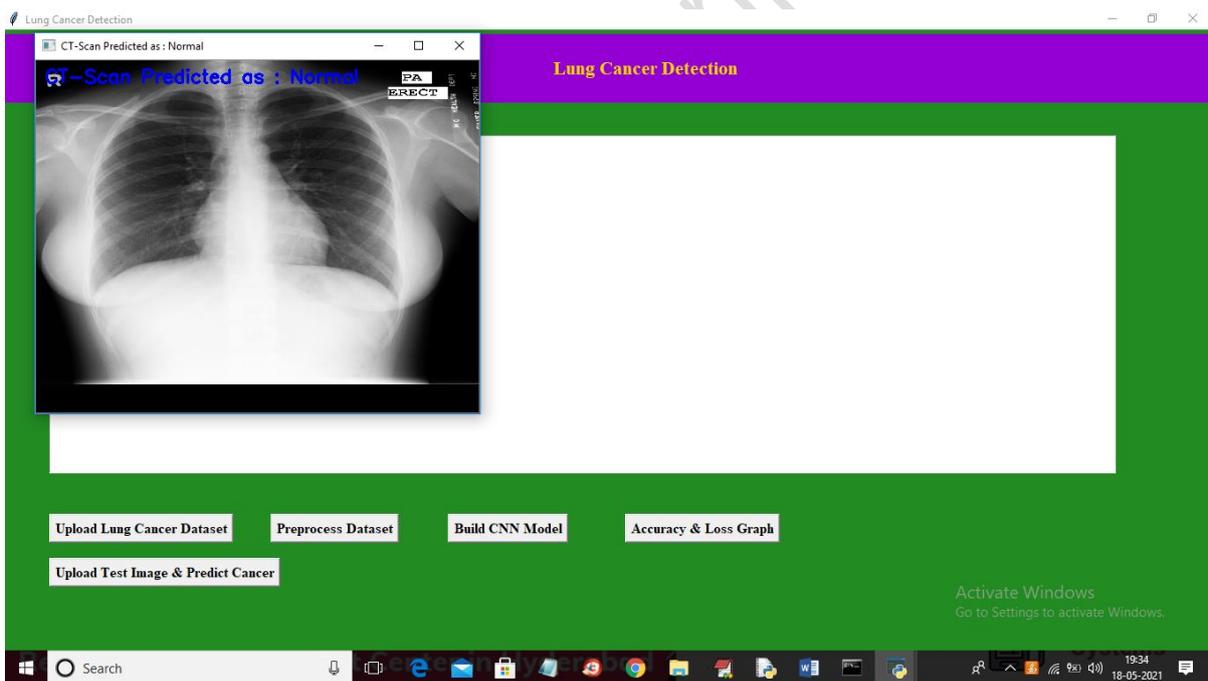


Fig 5:In above screen CT-SCAN is predicted as NORMAL. Similarly you can upload and test other images

5.CONCLUSION

In before times, the medical doctor has to do more than one exams in order to notice whether or not a given affected person has lung most cancers or now not .But this used to be a very time ingesting process. In a analysis once in a while a affected person has to endure useless check-ups or special assessments to perceive the disorder of lung cancer. To decrease the system time and needless check-ups there wants to be a preliminary check in which each the

affected person and the physician will be notified with the chances of lung cancer. Nowadays the computer gaining knowledge of algorithms performs an necessary position in the prediction and classification of clinical data. we can see envisioned end result as CT-SCAN consists of abnormality and in 2nd picture we are detecting locations have been abnormality detected and in 1/3 photograph we extracted all abnormality patches from authentic photograph and then displaying.

FUTURE SCOPE

As a future work, the experiments could be performed by using Deep CNN architecture for other types of cancer.

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