



International Journal of Innovative Research in Computer and Communication Engineering

(A Monthly, Peer Reviewed, Refereed, Scholarly Indexed, Open Access Journal)





Deep Learning Techniques for the Identification of Lung Cancer

Jyothi Dola¹, Govind Srivani², P.Sindhuja Reddy³

Department of AI & ML, CMR Institute of Technology, Hyderabad, India^{1,2,3}

ABSTRACT: Lung cancer is one of the world's most prevalent causes of passing away for both men and women. The WHO predicts that two million cases of lung cancer occur annually. If lung cancer is detected early, the average 5-year survival percentages for patients increases from 16 to 56%. A computed tomography (CT) scan can yield important data for lung disease diagnosis. The primary goals during this study are to identify lung cancer based on its severity and to detect cancerous lung nodules from the provided input lung image. To identify the cancerous lung nodules, this study uses state-of-the-art Deep learning algorithms. CT scanned lung pictures of cancer patients are acquired from a number of institutions. Areas of interest are separated using image processing techniques including pre-processing and segmentation techniques like feature extraction. Overall accuracy indicates the system is fairly effective at classifying instances in either positive or negative classifications. In addition, the use of a simple interface improves accessibility by enabling healthcare providers to submit CT scans for automated cancer prediction, which expedites prompt treatments and eventually improves the results for patients.

KEYWORDS: Deep learning, Convolutional Neural Network, ResNet

I. INTRODUCTION

According to the number of fatalities caused by lung cancer with regard to every other cancer types, lung cancer disease is the second leading cause of death around the world, after heart attacks. [1]. Cell proliferation without regulation that leads to the development of lung nodules is a feature of lung cancer. Approximately 19% of fatalities worldwide have been linked to lung cancer, primarily as a result of alcohol and tobacco use. Many techniques, such X-rays, Computed Tomography (CT), Magnetic Resonance Imaging (MRI), and Sputum Cytology, are employed to diagnose lung cancer. These techniques have an issue of being extremely time-consuming and only enabling detection at a later stage. Even though CT scans are the most noticeable imaging technology available, it can be difficult for doctors to interpret and identify cancer from the data. In order to correctly identify the cancerous cells, doctors may find computer-assisted diagnostics helpful. Computer-assisted techniques like image processing and deep learning have been used.

As cancer, it may access nearby tissues and proceed to other organs through blood arteries or lymph nodes at a higher level. Effective early detection and treatment of lung cancers depend on the quick identification as well as knowledge of the many cell types involved. Deep neural networks are implemented by DL to extract meaning from data. Because there are so a few indicators of lung cancer, early identification is difficult. Lung cancer has a chance to spread to other diseases. The term "metastasis" describes how cancer spreads across various body areas.

Malignant tumours, additionally referred to as malignant brain growths, are generally divided into two categories: primary tumours, which start in the brain tissue itself, and secondary tumours, which spread to the brain from other main places in the body. Exposure to chemicals such as vinyl chloride, neurofibromatosis symptoms, ionizing radiation exposure, and other environmental variables are among the risk factors linked to brain tumours. The study were designed to identify the most significant essential machine learning framework using data that reflected real-world scenarios. These findings give physicians important information and contribute to more accurate and timely lung nodule detection. The extensive use of CT image imaging requires the development of a filter to reduce noise, which alters the characteristics of CT pictures[2]. A few of the associated. The CT images have been divided using a variety of thresholding-based segmentation techniques. The majority of earlier studies are designed to evaluate various thresholding-based image segmentation algorithms according to attributes like accuracy, stability with regard to parameter selection, and stability with regard to image selection. Accuracy is measured using performance metrics such as precision, specificity, and false



International Journal of Innovative Research in Computer and Communication Engineering (IJIRCCCE)

(A Monthly, Peer Reviewed, Refereed, Scholarly Indexed, Open Access Journal)

positive rate. For the purpose of to precisely identify cancerous cells, doctors may find computer-assisted diagnostics useful. Computer-assisted methods such image processing and deep learning have been used. We have attempted to solve these issues in our proposed algorithm. The system we use is capable of identifying cancerous cells and each of their stages, including the beginning, middle, and the terminal stages. It determines the likelihood of lung cancer even though there are no cancer-affected cells in the input image.

II. RELATED WORK

Deep learning was recently recognized as a possible solution for enhancing lung cancer detection. Deep Learning models show the potential for improving the accuracy as well as efficiency of early detection methods through the use of intricate neural networks and advanced algorithms. This review of the literature examines how lung cancer identification is evolving, looking at major advances, difficulties, and the revolutionary effects of artificial intelligence in transforming the diagnostic process for this major medical issue. The possibility of computer-aided diagnostics in the medical field was proposed with J. Yanase and identifies in the year 2019. In this study, multiple recent research achievements in the direction of these issues are presented and certain research results have been discussed. The primary objective of M. Aharonuet al.'s study paper from 2023 included using DL techniques for detecting lung cancer. They conducted an extensive review of the current Deep Learning techniques for identifying lung cancer. In order to provide an accurate prognosis for lung cancer, V. Rajasekar et al. (2023) used deep learning and feature analysis of CT and histological images. It connects revealed an unconventional deep learning method that has been created for lung cancer testing using neural networks[3].

F. W. Alsaade with combines performed a study in 2021 that studied the use of artificial intelligence algorithms in the process of developing a lung lesion detection system for the aim of melanoma identification. The proposed solution was developed using both state-of-the-art deep learning algorithms and more traditional artificial intelligence machine learning methods[4].

Nisar et al. (2021) discussed the potential, hazards, and novel methods to deep learning in healthcare. The identification of breast cancer lymph node metastases, mental health issues, and other related illnesses are only a few of the numerous medical uses of deep learning that will be discussed in this article. This category includes problems of the heart, brain, and lungs. Following a summary of each group's findings, comparison tables are created using key criteria. DL models employ a wide range of resources, including data, software, hardware, and techniques. Finally, we discuss some of the current challenges facing deep learning models and where we believe future studies[5].

Having similar standards across healthcare platforms is necessary to provide smooth communication between various devices and systems, this is essential for the successful integration of deep learning models.

Furthermore, the ethical ramifications of using deep learning algorithms to make healthcare decisions raise questions about accountability, transparency, and the necessity of a human-centric approach, especially when dealing with potentially life-altering diagnoses like lung cancer. To fully utilize deep learning in an IoT healthcare setting for lung cancer detection, these problems must be resolved while giving patient privacy, data integrity, and ethical concerns first priority[6]. Numerous studies have been carried out on the subject of lung cancer detection; nevertheless, the standard research endeavour has been beset by inaccuracies and inefficiencies.

III. METHODOLOGY

1. Convolutional Neural Networks (CNN)

An artificial neural network having one or more convolution layers is called a convolutional neural network (CNN). CNN's are mainly employed for image processing, segmentation, classification, and other auto-correlated data. Deep learning is a machine learning-based artificial neural network that utilizes higher layers to gradually acquire characteristics from input in order to identify objects in image. As observed in the image, we have to provide the CNN with human faces in order to identify faces in images. CNNs provide an advantage of being able to create an internal representation of a two-dimensional image. These networks were extremely effective at extracting features that have been necessary for tasks like classifying images. The image below shows the essential functions of CNNs. Convolution is a basic technique in which this main layer processes elements from incoming images. Here, a channel with particular



International Journal of Innovative Research in Computer and Communication Engineering (IJIRCCCE)

(A Monthly, Peer Reviewed, Refereed, Scholarly Indexed, Open Access Journal)

dimensions (MxM) and the input image undergo numerical convolution to acquire information that enables feature extraction along with additional analysis. It is a neural network that is completely feed forward[7]. It was produced by the final few layers. A vector is generated after the image was convolved, pooled, and flattened. An ANN uses this vector as its input layer and functions normally to identify the object in the image. Each neural connection receives a random weight, and the input layer is subsequently weighed and fed into an activation function. Every and every neuron is connected to every other neuron in each subsequent layer. After an analysis of the output with the actual data, the error is back-propagated, indicating that all procedures are repeated and the weights are changed. This is carried out until the error decreases or the result is accurate.

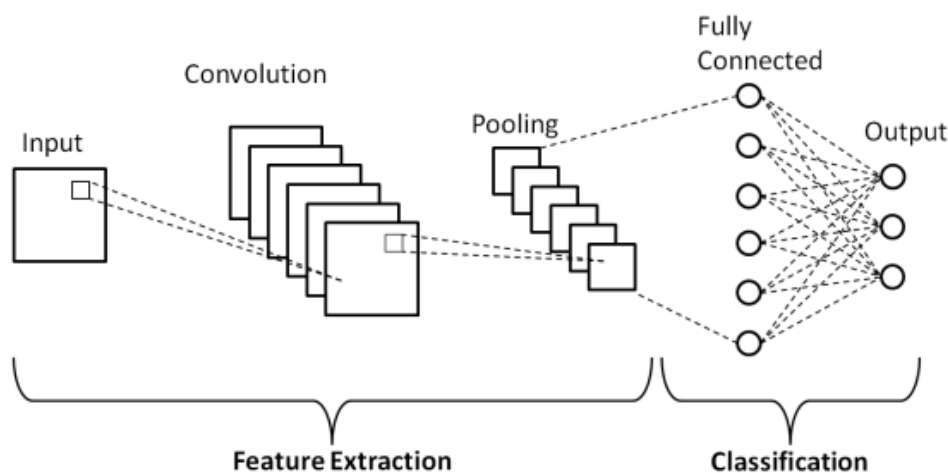


Figure 1: Architecture of CNN

1. Pooling or sub-sampling

The pooling method, frequently referred to as spatial pooling or sub-sampling, helps in lowering the feature maps' dimensionality while preserving the most important data. The 3D feature maps are transformed into a flattened one-dimensional feature vector by the method of pooling. Improving the accuracy and efficiency of planning algorithms needs this step.

2. Fully connected

The feed forward architecture of totally related layers in neural network structures is similar to that of the conventional neurons. These layers embody an essential component in the centre of the network structure, creating complex connections with all previous neural processes.

3. VGG-19 Model

By using its deep architecture, the VGG-19 model, a deep CNN, succeeds well in the classification of images related to lung cancer. VGG-19, featuring 19 layers, is effective at identifying variations in images of lung cancer given that it is capable of recording complex elements in images. The model's convolutional layers extract hierarchical features through a series of operations, and the fully connected layers help with the final classification[8]. By automatically learning and identifying complex patterns, textures, and statistical data from CT images, VGG-19 shows its prowess in classification of lung cancer. The model provides an excellent choice for accurate lung cancer image classification due to its deep architecture, which enables it to differentiate between true and false labels with accuracy.

In the final analysis, VGG-19 excels in classifying images of lung cancer since it can accurately identify between true and false labels by collecting and understanding complex elements from medical images. The model can be utilized for medical image analysis due to its deep approach which enables it to identify minor features that are essential for precise classification. Considering its unique capacity to collect detailed representations, this model is particularly appropriate to identifying minute patterns that may be signs of lung cancer.



International Journal of Innovative Research in Computer and Communication Engineering (IJIRCCCE)

(A Monthly, Peer Reviewed, Refereed, Scholarly Indexed, Open Access Journal)

These collected features are then used by the last fully linked layers to categorize the images into true or false labels[9].

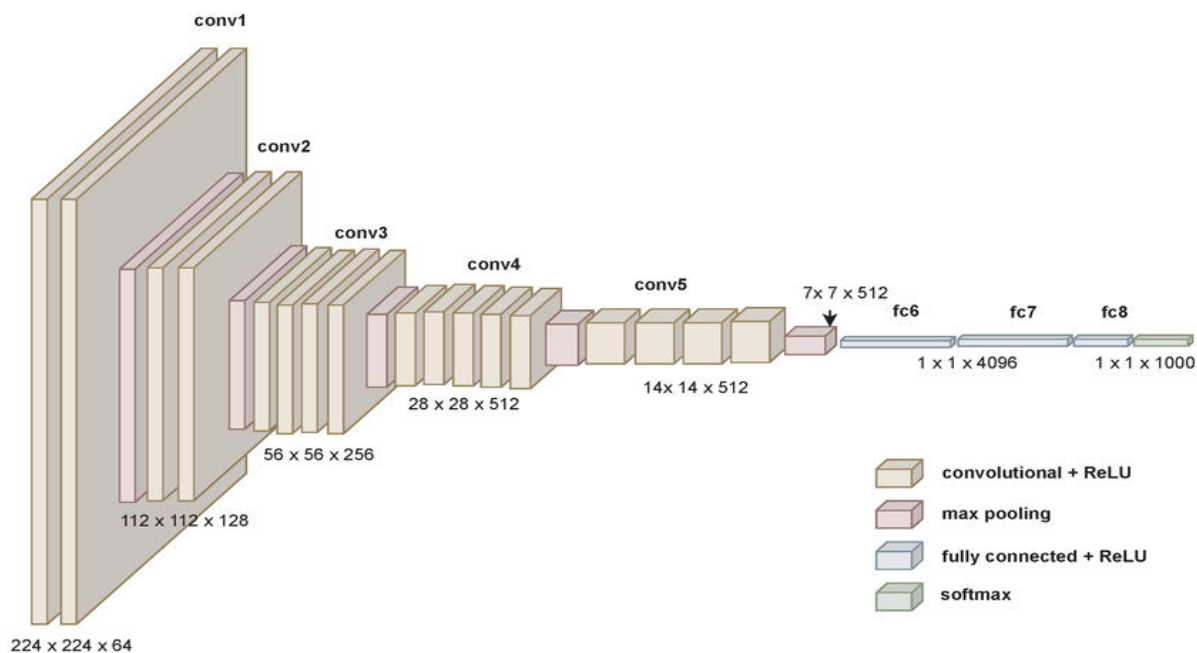


Figure 2: Architecture of VGG-19 model

4. ResNet

One of the different versions of the convolutional neural network ResNet is ResNet-50, featuring 50 layers. In addition to one MaxPool and one Average Pool layer, it has 48 Convolution layers[10]. When compared to time and memory, ResNet 50 performs more effectively than VGGNet. Computer vision models find it challenging to recognize masked faces since the features needed for accurately predicting the identity of an individual are reduced from all of the facial features to only their eyes occasionally the area around it. A ResNet-50-based architecture that works well at recognizing masked faces and networks with more layers can be trained easily without increasing the training error percentage.

This study utilizes an existing pretrained ResNet-50 architecture trained on human faces to tackle the problem of identifying a person's identity when wearing a face mask. It additionally helps with solving the vanishing gradient problem using identity mapping. The results of the research may be easily incorporated into current software for facial recognition, which is designed to recognize faces for security verification.

IV. RESULTS

We employed a dataset comprising 920 high-resolution images from the Kaggle platform to precisely train a Convolutional Neural Network (CNN) model. For the purpose to ensure optimal model performance, our methodology included stringent preprocessing procedures to improve the image quality and standardize features. By performing thorough testing and hyperparameter adjustment, our CNN model exhibits exceptional competence in precisely identifying possible lung cancer symptoms in radiographic images. Notably, our method enables early diagnosis by using the enormous potential of deep learning algorithms to identify complex patterns in medical images[11]. The system is able to distinguish between benign or non-cancerous nodules and cancerous or malignant nodules with a high degree of accuracy by training a CNN model on a dataset of CT lung images. The CNN is capable of recognizing tiny patterns indicating of cancer development through the use of many layers of convolutional and pooling methods.



International Journal of Innovative Research in Computer and Communication Engineering (IJIRCCCE)

(A Monthly, Peer Reviewed, Refereed, Scholarly Indexed, Open Access Journal)

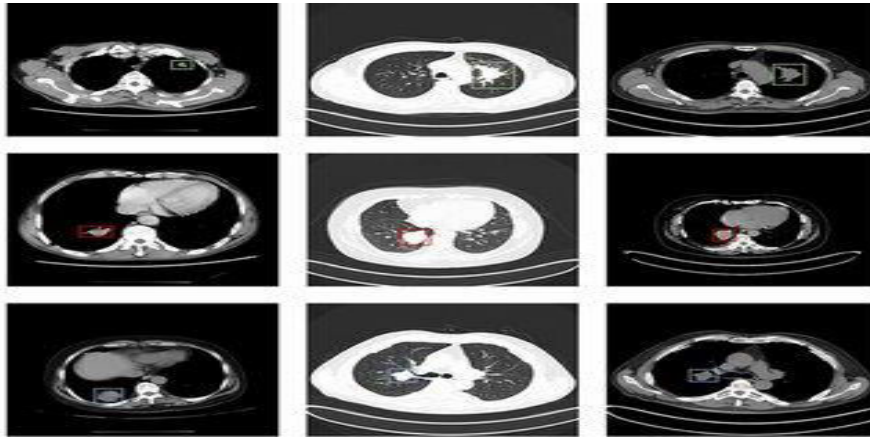


Figure 3: Lung Cancer Detection Image

According seen in Figure 3, the dataset used in this study includes 800 CT lung pictures, among which 563 are used for training and 237 for testing. These pictures are essential for both training and assessing how well the proposed deep learning model detects lung cancer. Each image in the set is a CT lung scan that captures various aspects of the lung's anatomy and potential Modules. For the purpose of to ensure optimal model performance, our technique included extensive preprocessing procedures to improve the image quality and standardize features. Through means of thorough testing and hyperparameter modification, our CNN model displays exceptional skill in precisely identifying possible lung cancer symptoms in radiographic images. Notably, our technique enables early diagnosis by using the enormous potential of deep learning algorithms to identify complex patterns in medical images.

Table 1: Comparison of The Research Deep Learning Models' Performance

| Model | Accuracy | Precision | Recall | F1-score |
|--------|----------|-----------|--------|----------|
| CNN | 0.88 | 0.89 | 0.90 | 0.89 |
| VGG-19 | 0.94 | 0.95 | 0.93 | 0.92 |
| ResNet | 0.90 | 0.92 | 0.90 | 0.91 |

The results show how deep learning could potentially used to improve detection systems. By automated the recognition of lung cancer in CT scans, hospitals along with additional healthcare facilities could promote their early detection efforts, improve diagnostic accuracy, and guarantee that patients suffering lung cancer receive appropriate treatment.

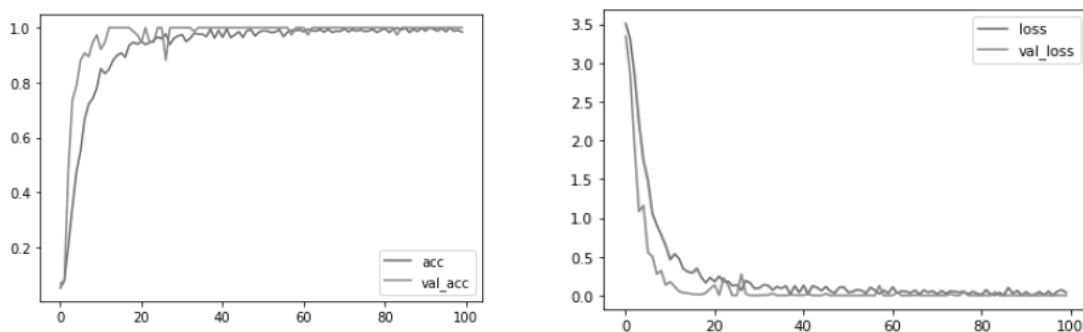


Figure 4: obtained CNN, VGG 16, and RestNet accuracy loss.



International Journal of Innovative Research in Computer and Communication Engineering (IJIRCCCE)

(A Monthly, Peer Reviewed, Refereed, Scholarly Indexed, Open Access Journal)

V. CONCLUSION

The present implementation improves previous initiatives by combining many approaches to accurately classify the condition. Using predicts from convolutional neural networks, we are able to identify cellular disintegration in the lung stage with high accuracy. The system effectively analyzes the affected region in lung images to detect early-stage cancerous tumors by combining CNN with modern methods of image processing which includes segmentation and feature extraction. The finding has implications for human health as it provides an advanced decision support system that can accelerate medical operations and promptly address concerns of patients. Automating the process of determining when lung nodules are cancerous. accessibility to early cancer diagnosis and accelerates the diagnostic procedure. The system's usability continues to be enhanced by the development of a user interface that is simple to use.

REFERENCES

1. Eali Stephen Neal Joshua, Midhun Chakkravarthy, Debnath Bhattacharyya, An Extensive Review on Lung Cancer Detection Using Machine Learning Techniques: A Systematic Study (Revue d'Intelligence Artificielle, Vol. 34, No. 3, 2020).
2. Gagan Thakral, Sapna Gambhir, Nagender Aneja, Proposed methodology for Early Detection of Lung ancer with low-dose CT scan using Machine Learning (International Conference on Machine Learning, Big Data, Cloud and Parallel Computing (COM-IT-CON, 2022).
3. V. Rajasekar et al, A. Ravi, T. Srikanth, B. Hari Praveen, h B. Sai Krishna, A.Sunil Mallik, Lung Cancer Detection using DL Approach CNN (International Conference on Communication and Electronics Systems (ICCES), 2023).
4. F. W. Alsaade, Sapna Gambhir, Nagender Aneja, Proposed methodology for Early Detection of Lung ancer with low-dose CT scan using Machine Learning (International Conference on Machine Learning, Big Data, Cloud and Parallel Computing (COM-IT-CON, 2021).
5. Nisar et al, Varshashree, Vansika Singh, Dheeraj, Donal Jovian Nazareth, Dr. Savitha Hiremath, A Web Application for Lung Cancer Analysis and Detection using Deep Learning Approach (International Journal for Research in Applied Science & Engineering Technology (IJRASET), Volume 10 Issue VI, 2021).
6. Mattakoyya Aharonu, R Lokesh Kumar, CNN based Framework for Automatic Lung Cancer Detection from Lung CT Images (International Conference on Smart Generation Computing, Communication and Networking (SMART GENCON), 2022).
7. M. Praveena, A. Ravi, T. Srikanth, B. Hari Praveen, h B. Sai Krishna, A.Sunil Mallik, Lung Cancer Detection using DL Approach CNN (International Conference on Communication and Electronics Systems (ICCES), 2022)
8. Amjad Khana, Zahid Ansari, Identification of Lung Cancer Using Convolutional Neural Networks Based Classification (Turkish Journal of Computer and Mathematics Education, Vol.12 No.10, 2021).
9. M. Praveena, A. Ravi, T. Srikanth, B. Hari Praveen, h B. Sai Krishna, A.Sunil Mallik, Lung Cancer Detection using DL Approach CNN (International Conference on Communication and Electronics Systems (ICCES), 2022).
10. Rohit Halder, Rajdeep Chatterjee "A deep learning based smart attendance monitoring system" uploaded by Rajdeep Chatterjee on 11 April 2020
11. Showkat A. Dara , and S.Palanivelb "Neural Networks (CNNs) and Vgg on Real Time Face Recognition System" Turkish Journal of Computer and Mathematics Education Vol.12 No.9 (2021), 1809- 1822.



INTERNATIONAL
STANDARD
SERIAL
NUMBER
INDIA



INTERNATIONAL JOURNAL OF INNOVATIVE RESEARCH

IN COMPUTER & COMMUNICATION ENGINEERING

 9940 572 462  6381 907 438  ijircce@gmail.com



www.ijircce.com

Scan to save the contact details