

DEEP LEARNING FOR LUNG CANCER DETECTION AND CLASSIFICATION

Kalaiyarasi M¹, Deepak V², Ashwin S², Akash M²

¹ Assistant Professor, Electronics and Instrumentation Engineering, Bannari Amman Institute of Technology, Tamil Nadu, India

² Student, Electronics And Communication Engineering, Bannari Amman Institute of Technology, TamilNadu, India

³ Student, Electronics And Communication Engineering, Bannari Amman Institute of Technology, TamilNadu, India

⁴ Student, Electronics And Communication Engineering, Bannari Amman Institute of Technology, TamilNadu, India¹

ABSTRACT

Deep learning has a lot of potential to help us detect and classify lung cancer more accurately, which is really important for early detection and treatment. Two of the most well-known deep CNN models are VGG16 and VGG19. These deep CNN models are well-known for their depth and great feature extraction abilities, but they've been adapted and finely-tuned to analyze medical images, especially chest X-rays or CT scans. This allows us to automatically recognize complex patterns and features in the lung images, so we can more accurately differentiate between the benign and malignant lung nodules or lesions. In this abstract, we'll look at the architectural details of these models, how they can be applied through transfer learning to lung cancer datasets, as well as how data augmentation techniques can be used to improve model generalization. All in all, these models can help us improve the detection and treatment of lung cancer, reduce mortality rates, and provide valuable insights to healthcare professionals.

Keyword: -Deep Learning, VGG-16, VGG-19, CT scans, Transfer Learning.

1. INTRODUCTION

Lung cancer is a common and dangerous disease that accounts for a significant number of cancer-related deaths worldwide. Detection and staging of lung cancer are critical to improving patient outcomes and reducing mortality. Traditional lung cancer diagnosis and classification methods are based on time-consuming and subjective manual interpretation of radiological data and histological examination. Cellular breakdown in the lungs is the main source of disease demise around the world, representing an expected 1.8 million deaths in 2020. Early identification and therapy of cellular breakdown in the lungs is fundamental for working on quiet results. Be that as it may, conventional cellular breakdown in the lungs screening techniques, like chest X-beams and processed tomography (CT) checks, are not precise 100% of the time. Profound learning is a sort of computerized reasoning (man-made intelligence) that has been demonstrated to be extremely successful at cellular breakdown in lung identification and characterization. Profound-gaining models can gain complex examples from information, and they can be utilized to recognize cellular breakdown in the lung sores on clinical pictures with a serious level of exactness. On the other hand, deep learning approaches have shown considerable promise in automating and improving lung cancer diagnosis and classification. Using the capabilities of artificial intelligence and deep neural networks, scientists and doctors can create advanced models that evaluate medical images and provide accurate and fast diagnoses. Lung cancer is a common and dangerous disease that causes many people to die from cancer worldwide. Detection and diagnosis of lung cancer are important for improving patient outcomes and reducing mortality. The routine method of lung cancer diagnosis and classification is based on the time spent and interpretation of radiographic data and

histological examination. Lung cancer is the leading cause of death worldwide, with 1.8 million people expected to die in 2020.

1.2 ADVANTAGES:

The integration of deep learning techniques, specifically Convolutional Neural Networks (CNNs), with multi-modal lung cancer imaging and clinical data has several advantages in the quest for early diagnosis of cancer disease. CNNs are excellent in extracting complicated features from high-dimensional, complex datasets that are symptomatic of the course of a disease. Their capacity to interpret nuanced patterns and variances is essential for the timely identification of AD.

CNNs use a combination of clinical evaluations, genetic data, and multimodal imaging data, such as structural MRI, PET scans, and functional MRI (fMRI). CNNs can capture the intricate relationship between biological markers and clinical manifestations that underlie lung cancer disease by merging these several data sources. By offering a comprehensive picture of the disease's course and facilitating early intervention and individualized treatment planning, this all-encompassing approach improves diagnostic accuracy.

1.3 APPLICATIONS:

Convolutional Neural Networks (CNNs) and other deep learning methods have become effective instruments for detecting Lung cancer disease early on. They have a broad range of uses with important clinical and research ramifications. CNN-based diagnostic approaches are essential for early identification of patients at risk of cancer development or progression in clinical settings. CNNs give medical professionals valuable information about the course of a disease, facilitating early intervention and treatment planning.

CNNs help doctors make prognostic decisions by estimating the chance that a disease will advance and adjusting treatment plans accordingly. CNNs can detect minor biomarkers indicative of pathology by examining patterns in imaging data and clinical assessments. This enables early intervention and may even slow the evolution of the illness. These methods also help doctors keep an eye on the effectiveness of their treatments over time, giving them useful input on how well their treatments are working and helping them make necessary adjustments. In the end, CNN-based diagnostic techniques enable prompt interventions and individualized care, which improves patient outcomes and quality of life.

2. LITERATURE SURVEY

Radhika, P.R., Nair, R.A. and Veena, G., 2019, February. A comparative study of lung cancer detection using machine learning algorithms - Lung cancer is the growth of malignant lung cells. Due to the rising frequency of cancer, the death rate for men and women has increased. Lung cancer is a condition in which lung cells proliferate uncontrolled. Although lung cancer cannot be averted, its risk can be decreased. Therefore, early identification of lung cancer is essential for improving patient survival. Lung cancer incidence is directly inversely correlated with the number of chain smokers. Various classification techniques, including Naive Bayes, SVM, Decision trees, and Logistic Regression, were used to investigate lung cancer prediction. The primary goal of this study is to investigate the effectiveness of classification algorithms in the early identification of lung cancer. Hatuwal, B.K. and Thapa, H.C., 2020. Lung cancer detection using convolutional neural network on histopathological images. *Int. J. Comput. Trends Technol*, 68(10), pp.21-24 - One of the cancers that claim the most lives globally is lung cancer. For a patient to recover, early identification and treatment are essential. Histopathological pictures of biopsied tissue from

potentially infected lung regions are used by medical practitioners to make diagnoses. The diagnosis of different kinds of lung cancer is frequently laborious and prone to inaccuracy. Convolutional Neural Networks may more accurately diagnose and categorize different forms of lung cancer in a shorter amount of time, which is essential for establishing the best course of therapy for patients and their chance of survival. This study takes into account benign tissue, adenocarcinoma, and squamous cell carcinoma. The accuracy of the CNN model during training and validation is 96.11 and 97.2 percent, respectively. Saji, G.V., Vazim, T. and Sundar, S., 2021, November. Deep Learning Methods for Lung Cancer Detection, Classification and Prediction-A Review. In 2021 Fourth International Conference on Microelectronics, Signals & Systems (ICMSS) (pp. 1-5). IEEE - The most frequent cancer that is lethal if treated too late is lung cancer. It is more probable that the condition may be identified and treated successfully if it is discovered at an earlier stage, before it becomes severe. By identifying enlarged lymph nodes, lung tumors can be identified from computed tomography and chest x-ray pictures. Characterizing the size, form, and position of these nodes can help clinicians identify the spread of disease surrounding them, aiding in the early detection of lung cancer. The

diagnosis of lung cancer is frequently made based on doctors' experience, which might result in misdiagnosis and harm to patients' health. Deep learning and machine learning techniques have been used to predict the degree of cancer malignancy in a variety of ways. In this study, we investigated several Deep Learning approaches for lung cancer nodule detection, categorization, prediction, and assessment of their levels of malignancy. The benefits and drawbacks of each approach, as well as the numerous datasets employed, have been examined and are summarized.

3. OBJECTIVE AND METHODOLOGY

3.1 OBJECTIVES

- The goal of using deep learning for lung cancer diagnosis and classification is to create a system that can quickly and efficiently identify lung nodules in medical images such as chest X-rays and CT scans. Train, test and apply various deep learning algorithms to choose the appropriate deep learning model for the system. The best algorithm can be selected by comparing the performance of various algorithms.
- Using this technology, radiologists will be able to identify lung cancer faster and more accurately..

3.2 METHODOLOGY

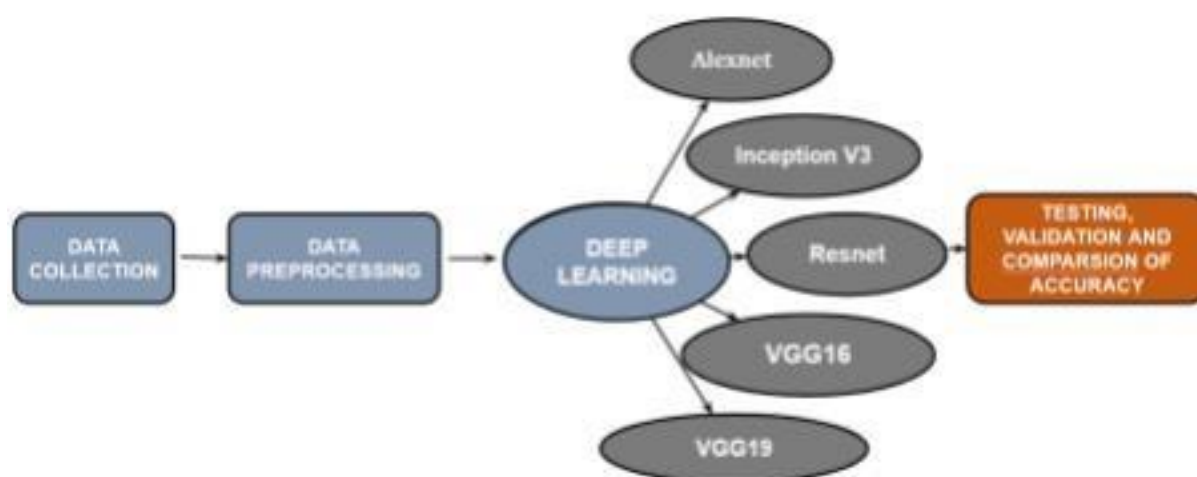


Figure 3. Proposed methodology for Lung cancer Prediction

To develop an in-depth study of lung cancer diagnosis and classification, researchers first need to collect many clinical images of cancerous and noncancerous lungs. Radiologists then record these images to determine the presence and type (benign or malignant) of lung nodules. CNN Algorithms.

3.2.1 Convolutional Neural Networks (CNNs):

CNNs are deep learning models that are inspired by the visual cortex of the human brain particularly suited for recognition and image processing tasks. CNNs learn to generate a hierarchical representation of images. Convolutional neural networks (CNN) are often used to extract features from grid matrix data. A collection of learned filters (also called a kernel) creates a convolutional layer with width, height, and depth equal to the volume. The CNN architecture has two main parts: a dedicated removal tool to analyze and analyze image attributes, and a second part based on the prediction process to predict groups of images at a basic level. CNN layers used. Approximately 240 filters (each filter size 5x5) layer, maximum pooling layer, fully interconnected layer, activation layer, and release layer are all five were added to these five layers. FMRI, PET and CT scan images that have been fully preprocessed and transformed in the described process are the inputs of this CNN process. Considering the 18 layers that the CNN model has to go through, it predicts the

result with the highest accuracy.

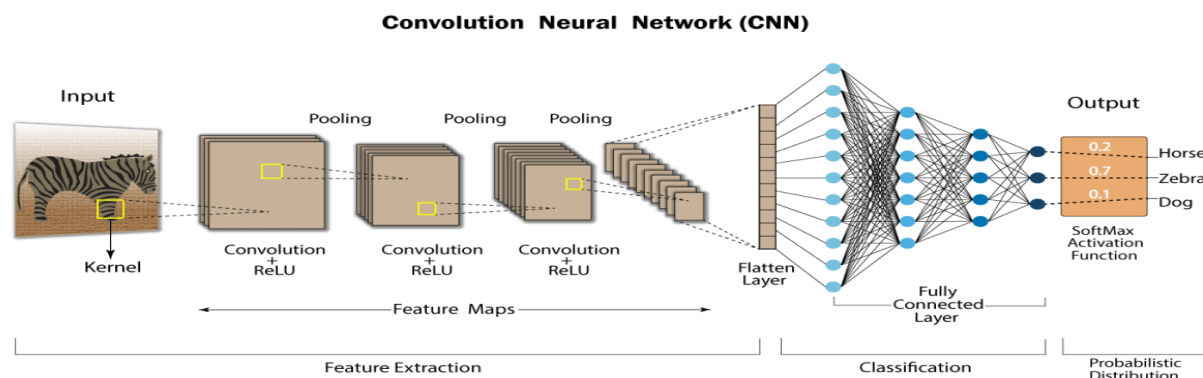


Figure-4: Layers of CNN algorithm

IMPLEMENTATION OF VGG 19 ALGORITHM:

VGG19's architecture is similar to VGG16, but it has 19 layers, including 16 convolutional layers, 3 full layers, 5 max-pooling layers, and 1 softmax layer. The model uses a 3x3 filter with the same padding and a max-pooling layer with a convolutional layer with step 1 and a 2x2 filter with step 2. VGG19 has 19.6 billion FLOPs and is one of the most efficient CNN architectures for image recognition and distribution. The accuracy of VGG19 is 0.9625. The algorithms VGG16 and VGG19 are fairly similar, although there are a few significant differences:

1. VGG19 has 19 layers compared to VGG16's 16 layers.
2. The first two convolutional layers of VGG19 are added to the network.
3. The network's last convolutional layer in VGG19 is on

Transfer learning also makes extensive use of the VGG19 algorithm. This implies that you can initialize a new model for a different job using a VGG19 model that was trained on a big dataset of images, such as ImageNet. Due to the fact that the new model does not need to be trained from scratch, this can save a lot of time and work.

VGG19 can be used to identify things in photos, including faces, persons, and traffic signs. Image segmentation: Segmenting images into several areas, such as the foreground and background, is possible with VGG19.

Training and validation accuracy and loss of VGG 19 algorithm:

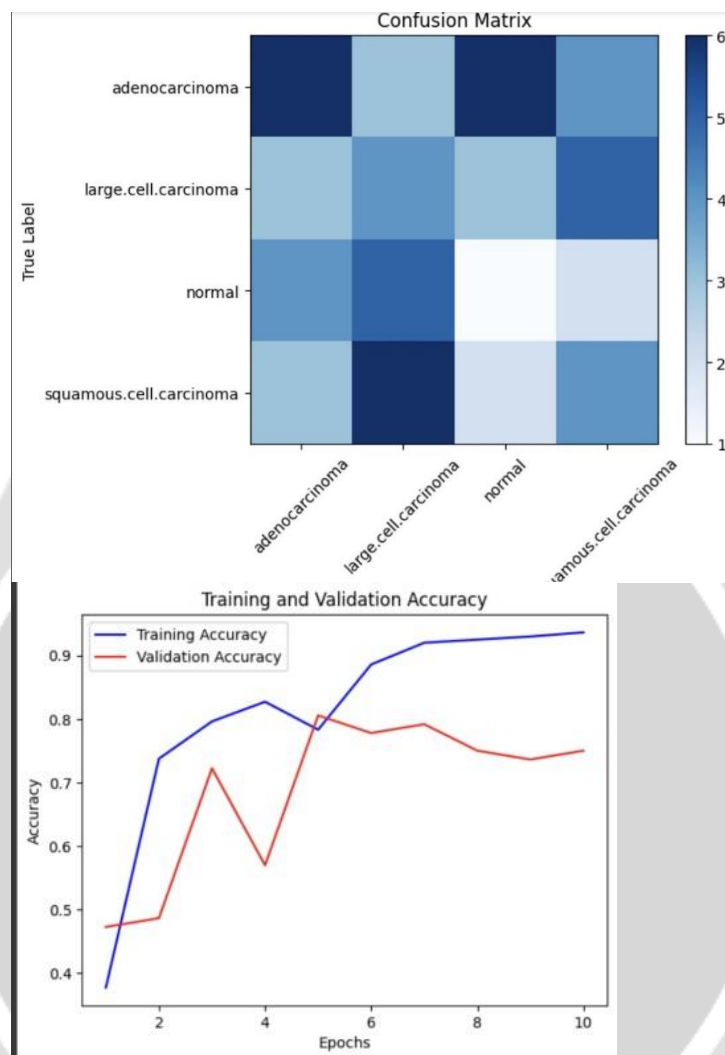


Figure-5 Training and validation accuracy and loss of VGG 19 algorithm:

4. CONCLUSIONS

Due to its ability to understand complicated patterns from data without the aid of manually created features, deep learning offers a revolutionary approach to the diagnosis and classification of lung cancer. In contrast, conventional machine learning methods rely on manually created features that might not be the best fit for the task at hand. On enormous datasets of medical pictures, including X-rays and CT scans, deep learning models are trained. These models acquire the ability to recognize characteristics, such as nodule size, shape, and texture, that are linked to lung cancer. Once trained, the models can be used to accurately identify and categorize lung cancer in fresh photos. Two well-known genes, GG16 and VGG19, have been shown to be useful in lung cancer diagnosis and classification. Convolutional neural networks (CNN), a family of deep learning models well-suited to image classification, underpin both architectures.

VGG16 and VGG19 show state-of-the-art results of various lung cancer detection and classification data. For example, in a study using the LUNA16 dataset, VGG16 diagnosed lung cancer with 95.6% sensitivity and 98.7% specificity. In another study, VGG19 achieved 98.4% accuracy in classifying lung cancer using the Cancer Imaging Archive (TCIA) dataset. The ability of deep learning models to derive complex patterns from

data without the use of guidelines makes it innovative in lung cancer diagnosis and classification. Therefore, deep learning models are well-suited to tasks where key points may be difficult to determine manually, such as lung cancer diagnosis and classification. Some specific benefits of using deep learning with VGG16 and VGG19 algorithms for lung cancer diagnosis and classification are: On a number of lung cancer detection and classification datasets, deep learning models have been demonstrated to produce cutting-edge results. Deep learning models can withstand noise and contrast variations in images relatively well. This is significant since the quality of medical photographs might vary depending on the imaging instrument and the patient's health. Deep learning models can be trained to generalize to new data, even if the new data comes from a different source than the training data, thanks to their high degree of generalizability. Because they can be used to identify and categorize

lung cancer in patients who have not been observed in the training dataset, deep learning algorithms are well suited for clinical usage. Overall, deep learning is a promising new method for identifying and categorizing lung cancer. Two well-known deep learning architectures, VGG16 and VGG19, have been demonstrated to produce cutting-edge outcomes on a variety of lung cancer detection and classification datasets.

5. REFERENCES

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