



# DETECTION OF COVID-19 AND PNEUMONIA USING CHEST X-RAY

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**Abstract:** Severe Acute Respiratory Syndrome Corona virus-2 is the cause of COVID-19. It is a contagious illness that spreads through tiny droplets of saliva or bodily fluid from an ill person's respiratory system when they cough, snuff, or hack. HIV spreads quickly through close contact with infected individuals or by touching, conserving, or touching surfaces and things exposed to the virus. Another contagious illness known as pneumonia is frequently brought on by contamination brought on by a bacterium in the lungs' alveoli. When a lung tissue that has been infected aggravates, it produces discharge. Professionals perform physical exams and use chest X-rays and lung ultrasounds to diagnose patients to determine whether they have these conditions. In this research, we create a method to recognise pneumonia and Covid-19 infection. The four conditions evaluated were pneumonia, coronavirus pneumonia, non-coronavirus pneumonia, and regular lungs. The proposed framework for computer-based intelligence is divided into two parts. Chest X-ray volumes are classified into pneumonia and non-pneumonia in stage 1. If X-Ray has a spot with pneumonic class and further ranks it into Coronavirus negative and Coronavirus positive, stage 2 receives a contribution from stage 1. Convolutional neural networks (CNNs) in particular have achieved successful outcomes in the categorization and analysis of medical image data using artificial intelligence (AI) approaches. This research proposes a deep CNN architecture for the classification of chest X-ray images in the diagnosis of COVID-19. An efficient and precise CNN classification was difficult due to the need for a chest X-ray image collection that was large enough and of high enough quality. The dataset has been pre-processed in different phases using different techniques to create a practical training dataset for the proposed CNN model to achieve its best performance. This was done to deal with these complexities, such as the availability of a very-small-sized and imbalanced dataset with image-quality issues. Pre-processing of the datasets used in this investigation involved dataset balance, picture interpretation by medical professionals, and data augmentation. The testing findings revealed an overall accuracy of 99.5%, demonstrating the suggested CNN model's strong suit in the relevant application domain. Two situations were used to evaluate the CNN model. Using the 100 X-ray pictures from the original processed dataset, the model was assessed on the first scenario and showed 100% accuracy. The model has been tested using a separate dataset of COVID-19 X-ray pictures in the second scenario. In this test situation, the performance reached 99.5%. A comparison study of the proposed model and several machine learning algorithms has been conducted to demonstrate further that it performs better than other models. When the proposed model was tested using an independent testing set, it outperformed all other models, both generally and specifically.

*IndexTerms:* convolutional neural networks (CNNs), Artificial intelligence (AI)

## I.INTRODUCTION

The quick spread of Coronavirus, thus the new SARSCOV-2 infection, is the most concerning issue confronting humankind today. Around one-half million patients on the planet kicked the bucket from Coronavirus this year (2020). Like this, it has gotten essential to recognize positive cases as fast as conceivable to forestall the further spread of this pestilence. Computer-based intelligence-based X-Beam screening is considered a promising methodology to test Coronavirus in asymptomatic patients. What's more, the location of Coronavirus in chest X-beam pictures is a complicated undertaking relying upon the presence of accomplished radiologists. Having a radiologist doesn't take care of the issue since the appearance isn't explicit and regularly equivocal, which prompts enormous contrasts between the radiologist during finding.

Pneumonia can be a problematic ailment if not analysed as expected and can bring about the demise of an individual related to this sort of sickness. It's anything, but extreme respiratory sickness brought about by infectious specialists like infections or microscopic organisms that influence the lungs. It tends to be spread through the nose or throat and impacts the lungs if they are breathed in or conveyed through airborne beads from an individual hacking or sniffing. An individual's lungs are comprised of little sacs or alveoli that provisions air entry at whatever point an all-around fit individual inhales. When an individual is contaminated with pneumonia, it restricts oxygen admission and makes breathing troublesome and difficult because of tissue touchiness brought about by alveoli covered with liquids or discharge. Maturing individuals 50 years old or more and children under five years old are powerless against pneumonia sickness for they have a more fragile resistant framework, and it has assumed control over 1,000,000 lives universally. The nation revealed almost 58,000 mortalities in 2016 and is the third top executioner behind heart infections and disease.

Coronavirus signs are practically indistinguishable from pneumonia and, if not appropriately analysed, will prompt mistaken analysis since numerous clinics all throughout the planet are clogged. Many of these clinics are working all day because of the monstrous increment of contaminations, and the greater part of its clinical staff is likewise tainted with the infection. Loose discoveries of pneumonia or non-Coronavirus might be marked mistakenly as Coronavirus tainted, and misfortunes of inappropriate treatment are outrageous, the battle and hazard of being presented to other positive patients of Coronavirus. The radiologist catches a chest picture of the patient through a radiograph instrument. A radiograph picture is produced through radiation on a sensitive film to approve whether patients are tainted with sickness or not contaminated. Even though standard CXR pictures could help early indications of suspected cases, the images of other related viral pneumonia are comparable and interrelate with other infectious lung sicknesses. Subsequently, for a radiologist, it's challenging to recognize Coronavirus to other related viral pneumonia.

## II.LITERATURE SURVEY

R. Sethi et al. [5] analyse the existing body of literature and the method of diagnosing diseases, stating that the RT-PCR test is the real-time standard diagnostic technique. The primary purpose is to find the COVID-19 virus or any other pathogen. The first step in the procedure is the collection of a nasopharyngeal swab from the patient's throat or nose. During the pandemic, it aids in the diagnosis of COVID-19 by the medical community. Cons of the test include the length of time it takes to receive results, incorrect reports, and inconvenience to patients during the test process. Using X-ray pictures as an alternate screening method will ease the burden of diagnosis on the doctor. Convolution neural networks (CNN) using deep learning techniques are used to classify X-Ray images. In order to train the CNN model to diagnose 14 diseases, around 100,000 chest X-ray images of various lung ailments were used. However, in order to compare the accuracy, seven distinct existing designs are used. Finally, fine-tuning increases the accuracy of the best model. For COVID-19 and other disorders, new designs based on the CNN detection model are presented. CNN model [6] can also be used to identify diabetics.

The complete blood counts (CBCs) of COVID-19 hematologic were analysed by T. Xia et al. As we know, Red blood cells (RBC), white blood cells (WBC), and blood platelets are all included in complete blood counts (CBCs). Anaemia, haemophilia, infection, or tumours are identified by a decrease in RBC, WBC, platelets, and blood cells with unique forms. Costly haematology of blood cell counts is time-consuming, but more straightforward, faster, and cheaper methods are available. YOLOv3 is the real-time object detection algorithm. CNN Darknet-53 and YOLOv3 introduce the brand-new feature. Three hundred sixty-four photos comprise the dataset utilised to study microscopic blood cells. They are training a particular classifier to recognise performance gradients displayed as bogus YOLOv3 weights. Finding blood cell samples with COVID-19 symptoms or blood clots, or any other COVID-19 symptoms is one of the benefits. The next step is to use an ensemble method to increase the test findings' recall and precision. A Point of Care (POC) device is used in the COVID-19 epidemic to aid in early detection and control efforts. An efficient way to employ COVID-19 detection is through a mobile application with microfluidic and microscopy access [7].

To track the patient's circumstances and to ensure prompt treatment, L. Xu et al. proposed lung lesion identification using CT images in RCNN. K-Means algorithm clusters pictures, and faster RCNN models utilising model VGG-16 and ResNet 50 classification are compared. The mentioned algorithm consists of two steps, the first of which obtains anchors from the input images by extracting features, and the second turns the acquired characteristics into pooling layers. They integrate the classification models at the end. ResNet50 may produce results within the original dataset comparable to those of VGG-16, but when the model is trained with an augmented dataset, it is more prone to overfitting. The current state of affairs requires that any model's accuracy be increased. So, for datasets with small image sizes, Faster CNN's accuracy must be enhanced, which will improve performance. They continue to work on enhancing the model structure and detection accuracy [9].

[10] Using COVID-19 radiological pictures, Z. Muftuoglu et al. demonstrated a significant application of data collecting, data size, and data quantity. The idea of differential privacy (DP) practice for such apps to be trustworthy in real-life use cases was the study's most significant contribution. To guarantee privacy, experiments were conducted with DP-applied photos and a private aggregation of teacher ensembles technique. There are some downsides to COVID-19 test scenarios. It displays falsely favourable findings and takes a short while to produce them. Additionally, it makes use of expensive machinery and infrastructure. Yet X-Ray imaging can get around these issues.

### III.NEED FOR STUDY

The COVID-19 pandemic is still having a terrible impact on the health and happiness of people worldwide. Effective patient screening, including radiological examination employing chest radiography as one of the primary screening modalities, is an important step in the fight against COVID-19. It was shown in early research that patients present anomalies in chest radiography pictures indicative of those infected with COVID-19. This paper introduces COVID-Net, a deep convolutional neural network design optimised for detecting COVID-19 instances from chest X-ray (CXR) pictures that are open source and accessible to the public. It was motivated by this and inspired by the open-source initiatives of the research community.

At the time of its original release, COVID-Net was, to the best of the authors' knowledge, one of the first open-source network architectures for COVID-19 identification from CXR pictures. We also provide COVIDx, an open-access benchmark dataset we created with 13,975 CXR images over 13,870 patient cases and, to the authors' knowledge, the most significant number of publicly available COVID-19-positive cases. Additionally, we look into how COVID-Net makes predictions using an explainability method to gain more insight into the crucial elements connected to COVID cases and help clinicians improve screening. We also audit COVID-Net responsibly and transparently to confirm that it makes decisions based on pertinent data from the CXR images. The open-access COVID-Net is by no means a production-ready solution. Still, it is hoped that researchers and citizen data scientists will use it and build upon it to speed the creation of highly accurate yet useful deep-learning solutions for detecting COVID-19 cases and expediting treatment for those who need it the most.

### IV.PROBLEM STATEMENT

The goal is to use chest X-ray pictures and CNN to detect COVID-19 and pneumonia-infected patients. The identification falls within the category of

- Identification of affected lungs
- Identification of Pneumonia affected lungs.
- Identification of COVID-19 affected lungs.

### V. IMPLEMENTATION

The real power of partitioning comes if a system is partitioned into modules so that the modules are solvable and modifiable separately. It will be even better if the modules are also separately compatible (then, changes in a module will not require recompilation of the whole system). A system is considered modular if it consists of discrete components so that each element can be implemented separately. Any change to one component has minimal impact on the other components.

Modularity is a desirable property in a system. Modularity helps in system debugging. Isolating the system problem to a component is easier if the system is modular. In system repair, hanging a part of the system is easy, as it affects a few other features, and in system building, a modular system can be quickly built by “putting its modules together.”

**Dataset Collection:** Kaggle chest X-ray info is an unbelievably widespread database with 15,798 chest X-ray pictures of traditional or healthy, viral, and bacterial pneumonia starting from 800 pixels to 1900 pixels. For the entire 15,798 image datasets, 5,396 pictures are units full of microorganism pneumonia and four,865 pictures with virus infection, and 5,537 images are units from traditional or healthy chest X-rays. Positive and suspected COVID-19 views were nonheritable in publicly available sources. Chest X-ray pictures for conventional and affected with respiratory disease were utilised from this assortment to generate the newest info assortment.

**Convolutional Layer:** This layer paperwork a basic Constructing block for convolutional neural networks. This layer makes use of a hard and fast size filter to extract numerous functions. The Inspection of photographs is achieved by means of shifting the filters according to Strides; in this example, there are six convolutional layers with Lengths of 32, sixty-four, sixty-four, 128, 128, and 128 filters in the CNN version. Every layer uses 2nd convolutional filters with an approximate size of 3x3 and a stride of 1.

**Batch Normalization:** It is wont to improve the Mastering charge of the CNN model, and this sediment standardizes the input photograph. Batch normalisation in an exceedingly CNN model is applied once every convolutional layer.

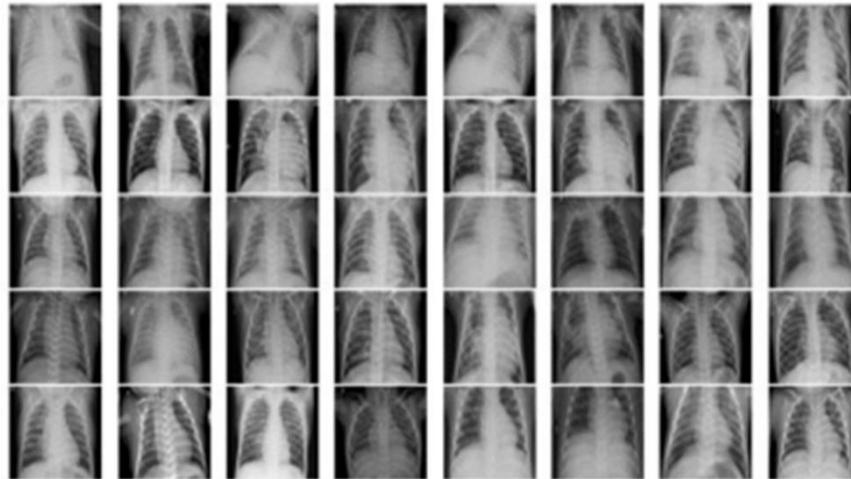
**Pooling Layer:** Pooling is a methodology that samples the amassed feature map from a convolutional layer. Max pooling and average pooling are commonly used, and in each Convolutional layer, a max pooling with a pooling filter out-length of 2\*2 is applied.

**Activation:** This characteristic is a non-linear Transformation of inputs that can be carried out at every cease of a layer. RLU or rectified linear unit is a not unusual activation function. This is applied at every cease of the layer, and within the very last layer, there are two nodes used with an activation performed.

**Dropout:** A technique applied to cut back the overfitting of the model. Bound nodes within the layer victimization, the dropout method is indiscriminately selected to be inactive occasionally. This can stop the model from obtaining excessively conversant in info—the dropout of zero. Five were employed within the dense layers of the model for classification.

**Dense Layers:** The output of the convolutional layer is, in addition, flattened and submitted as input to the dense layer. The convolutional-layer task is to extract features, and the role of the dense layer is for the classification of pictures. The CNN structure has two dense layers with 512 nodes every and a couple of nodes for the ultimate layer.

**Recognize:** From the data and performance metrics presented above, it can be seen that the best-performing model is the CNN trained for 100 epochs with a batch size of approximately 32 with an average accuracy of about 80% across the four classes: Covid19, Lung Opacity, Normal, and Viral Pneumonia. Even though the model does not classify all Covid-19 samples correctly, its accuracy matches that of radiologists.



Chest X-Ray Images

## VI. RESULTS AND DISCUSSION

For training, the architecture techniques like distributed computing, multiprocessing and parallel programming were used. During testing on X-ray volumes, we attained a maximum accuracy of 97%. The entire architecture was trained in Windows 10 Operating System. The software development language used is Python 3.6 version. The entire model is coded in the TensorFlow framework to train and develop the model using Kera's. The system used is Intel Core i5 - 7th Gen, 8GB RAM. The development of this architecture was a challenging task. So, to overcome these challenges, various methods were used like data augmentation, image-rotation, image-flipping, shift range, and zoom range. The model proposed in this research for pneumonia and Covid-19 detection on chest X-Ray images is VGG-16 using the CNN method. The result of this research shows a positive and accurate role of the Artificial Intelligence algorithm in detecting and identifying covid-19 patients very quickly. The medical field is the essential area that will get benefited from this research. In a future study, a more optimized and more developed complex network-structure version of this architecture is possible, which might be able to determine the best model for the detection of Covid-19 and pneumonia.

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