

#### PANIMALAR ENGINEERING COLLEGE





# DIAGNOSIS OF PNEUMOTHORAX USING DEEP LEARNING

Batch No: 24

#### **Batch Members:**

Alex Moses Inbaraj.A (211418107004)

Dinesh.R (211418107026)

Gnana Sai.A.R (211418107029)

#### **Internal Guide:**

Dr. R. Manikandan, M.E.Phd,

Professor,

Department of EIE,

Panimalar Engineering College



## **ORGANIZATION**



- ➤ Objective of the Project
- > Introduction
- Existing System/ Literature Survey
- Problem Identification
- Proposed system Block Diagram
- ➤ Algorithm Explanation
- Software Requirements
- > Experimental Process
- > Results & Discussions
- Conclusion And Future Work
- References



## **OBJECTIVE OF THE PROJECT**



To identify Pneumothorax by Deep Learning

Technique with the help of LENET and

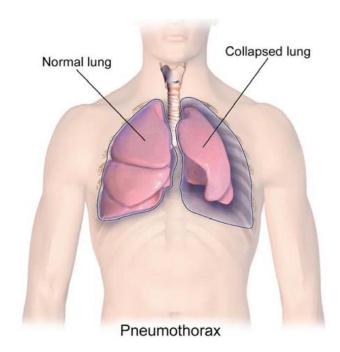
ALEXNET of CNN Architecture.



### **INTRODUCTION**



A pneumothorax is an abnormal collection of air in the <u>pleural</u> space between the <u>lung</u> and the <u>chest wall</u>. Symptoms typically include sudden onset of sharp, one-sided <u>chest</u> <u>pain</u> and <u>shortness of breath</u>.

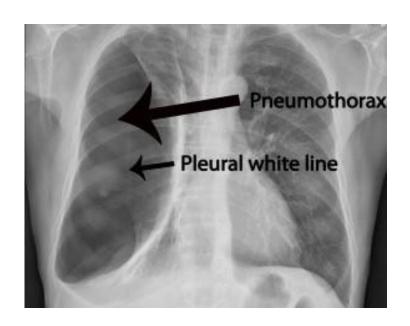




## **INTRODUCTION**



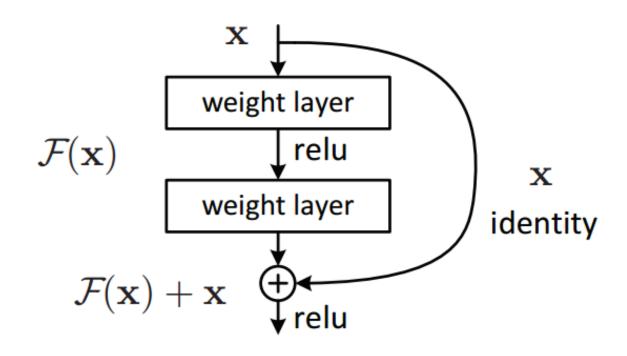
➤In a minority of cases, a one-way valve is formed by an area of damaged <u>tissue</u>, and the amount of air in the space between chest wall and lungs increases; this is called a tension pneumothorax.





## **BLOCK DIAGRAM OF PROPOSED SYSTEM**







# LITERATURE SURVEY

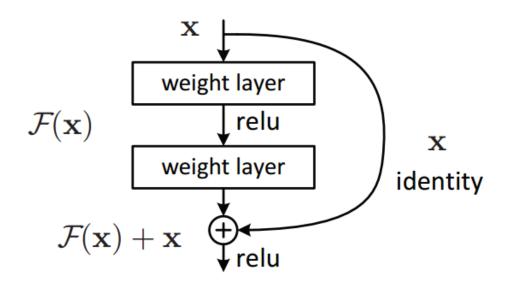


Author & Year	Title	Algorithms Used/ Achievements	Limitations
ErdiÇallı, EcemSogancioglu, Bramvan Ginneken,Kicky G.van Leeuwen, KeelinMurphy(2021)	Deep learning for chest X-ray analysis: A survey	Using deep learning on chest radiographs published before March 2021, categorizing works by task: image-level prediction (classification and regression), segmentation, localization, image generation and domain adaptation.	model may be complex and also inherits the disadvantages of wrapper model
Priyanka Malhotra , Sheifali Gupta , Deepika Koundal , Atef Zaguia , Manjit Kaur and Heung- No Lee.(2022)	Deep Learning- Based Computer- Aided Pneumothorax Detection Using Chest X-ray Images	The present work compares the operation of the proposed MRCNN model based on ResNet101 as an FPN with the conventional model based on ResNet50 as an FPN. The proposed model had lower class loss, bounding box loss, and mask loss as compared to the conventional model based on ResNet50 as an FPN	More Error rate



## **BLOCK DIAGRAM OF EXISTING SYSTEM**







## PROBLEM IDENTIFICATION/ LIMITATIONS OF EXISTING SYSTEM

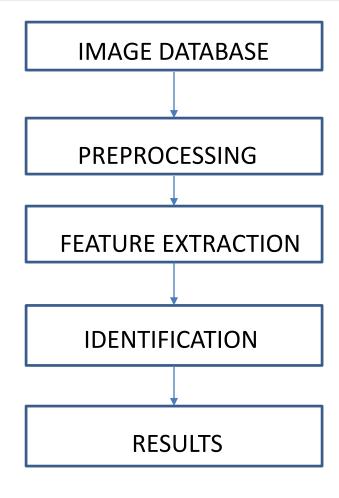


- Lesser efficiency.
- ➤ High run-time.
- ➤ Complex in process.
- The Error rate in RESNet-50 is more.



## **OVERVIEW OF THE PROPOSED WORK**





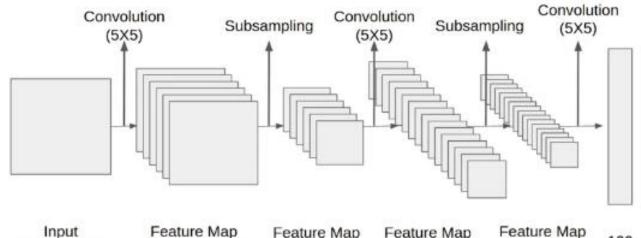


## **BLOCK DIAGRAM OF PROPOSED SYSTEM**



#### AlexNet Block Diagram





Feature Map 28 X 28 X 6 Feature Map 14 X 14 X 6 Feature Map 10 X 10 X 16 Feature Map 5 X 5 X 16

120



# BLOCK DIAGRAM OF PROPOSED SYSTEM OF



LeNet	AlexNet	
Image: 28 (height) × 28 (width) × 1 (channel)	Image: 224 (height) × 224 (width) × 3 (channels)	
<b>↓</b>		
Convolution with 5×5 kernel+2 padding:28×28×6	Convolution with 11×11 kernel+4 stride:54×54×96	
↓sigmoid	√ReLu	
Pool with 2×2 average kernel+2 stride: 14×14×6	Pool with 3×3 max. kernel+2 stride: 26×26×96	
<u> </u>		
Convolution with 5×5 kernel (no pad): 10×10×16	Convolution with 5×5 kernel+2 pad:26×26×256	
↓sig moid	√ReLu	
Pool with 2×2 average kernel+2 stride: 5×5×16	Pool with 3×3 max. kernel+2 stride: 12×12×256	
√ flatte n		
Dense: 120 fully connected neurons	Convolution with 3×3 kernel+1 pad:12×12×384	
↓sigmoid	√ReLu	
Dense: 84 fully connected neurons	Convolution with 3×3 kernel+1 pad:12×12×384	
↓sigmoid	√ReLu	
Dense: 10 fully connected neurons	Convolution with 3×3 kernel+1 pad:12×12×256	
<b>↓</b>	√ReLu	
Output: 1 of 10 classes	Pool with 3×3 max. kernel+2 stride: 5×5×256	
	√ flatte n	
	Dense: 4096 fully connected neurons	
	√ ReLu, dropout p=0.5	
	Dense: 4096 fully connected neurons	
	√ ReLu. dropout p=0.5	
	Dense: 1000 fully connected neurons	
	Output: 1 of 1000 classes	



#### PROPOSED SYSTEM



- We are proposing recognition framework based on the structured **two dimensional** Convolutional Neural Networks (CNNs) type of AlexNet and LENET to identify the Pneumothorax and improve the accuracy of workflow.
- The proposed method for this project is to train a Deep Learning algorithm capable of identifying and data preprocessing and visualizing the image then feature extracting to build **LENET CNN** using Pneumothorax image dataset. we classify it such as Pneumothorax or Normal this system using CNN model.
- It is predicted that the success of the obtained results will increase if the CNN method is supported by adding extra feature extraction methods and classify successfully Pneumothorax. We have demonstrated the efficacy and potential of using deep CNN to images.



# **ALGORITHM EXPLANATION**



- Lenet is a convolutional neural network that is 7 layers.
- You can load a pretrained version of the network trained images from the ImageNet database.
- ➤ The pretrained network can Identify images of 1000 CT scan and can show whether it is Pneumothorax diesease or normal.
- AS a result, the network has learned rich feature representations for a wide range of images.
- ➤ The network has an image input size of 225-by-225



## **SOFTWARE REQUIRMENTS**



#### **HARDWARE REQUIREMENT:**

> Processor: Pentium Dual Core 2.3 GHz Processor

➤ Hard Disk: 250 GB or Higher

> RAM: 2 GB (Minimum)

#### **SOFTWARE REQUIREMENT:**

Operating System: Windows 7 or Higher

Languages used: Python (OpenCV and CNN)

> Tools: Anaconda, Jupyter Notebook, Pycharm



## **EXPERIMENTAL PROCESS**



#### STAGE 1:

To increase the accuracy, We trained with number of images with ALEXNET model.

**For Training**:

**Dataset: 917** images

**For Testing:** 

**Dataset: 217**images

The Accuracy for images is 0.60



## **EXPERIMENTAL PROCESS**



#### STAGE 2:

To increase the accuracy, We trained with number of images with LENET-5 model.

**For Training:** 

Dataset: 917 images

**For Testing**:

**Dataset: 217**images

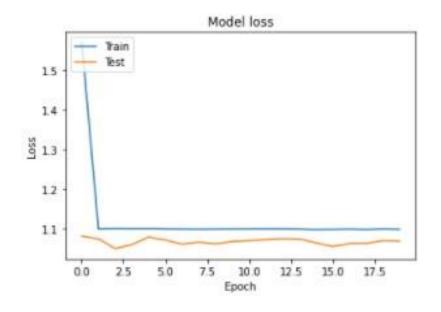
The Accuracy for images is 0.90

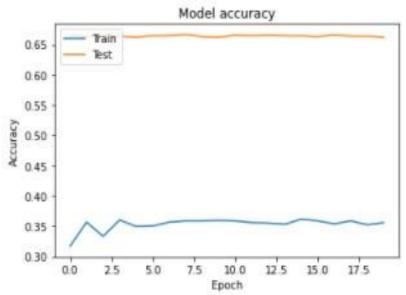




#### GRAPHS FOR TRAINED ALEXNET MODEL

:

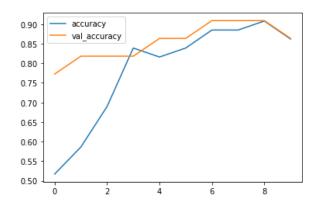


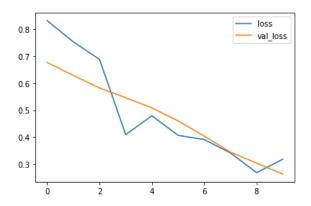






#### **GRAPHS FOR TRAINED LENET MODEL:**









#### **FOR MORE IMAGES:**

#### **Comparison of Evaluation Matrices:**

ARCHITECTURE	AUC	Sp(%)	Acc(%)
ALEX NET	58	60.01	61.08
LENET	<u>90</u>	89.03	88.02

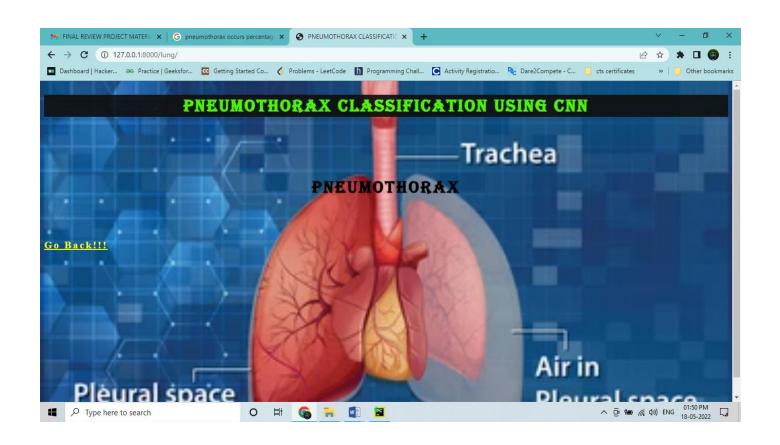
#### **Advantages:**

- The large amount of chest x-ray data can be train on artificial neural network.
- It is best model for deep learning technique to easily identifying Pneumothorax





#### **OUTPUT IMAGE:**





## REFERENCES



- [1]L. Thomsen, O. Natho, U. Feigen, U. Schulz, and D. Kivelitz, "Value of Digital Radiography in Expiration in Detection of Pneumothorax," RöFo Fortschritte Auf Dem Geb. Röntgenstrahlen Bildgeb. Verfahr., vol. 186, no. 03, pp. 267–273, Mar. 2014, doi: 10.1055/s-0033-1350566.
- [2] S. Sanada, K. Doi, and H. MacMahon, "Image feature analysis and computer-aided diagnosis in digital radiography: automated detection of pneumothorax in chest images," Med. Phys., vol. 19, no. 5, pp. 1153–1160, Oct. 1992, doi: 10.1118/1.596790.
- [3] O. Geva, G. Zimmerman-Moreno, S. Lieberman, E. Konen, and H. Greenspan, "Pneumothorax detection in chest radiographs using local and global texture signatures," in Medical Imaging 2015: Computer-Aided Diagnosis, Mar. 2015, vol. 9414, p. 94141P, doi: 10.1117/12.2083128.
- [4] Y.-H. Chan, Y.-Z. Zeng, H.-C. Wu, M.-C. Wu, and H.-M. Sun, "Effective Pneumothorax Detection for Chest X-Ray Images Using Local Binary Pattern and Support Vector Machine," J. Healthc. Eng., vol. 2018, p. 2908517, 2018, doi: 10.1155/2018/2908517.
- [5] A. Gooßen et al., "Deep Learning for Pneumothorax Detection and Localization in Chest Radiographs," ArXiv, 2019.
- [6] G. Kitamura and C. Deible, "Retraining an open-source pneumothorax detecting machine learning algorithm for improved performance to medical images," Clin. Imaging, vol. 61, pp. 15–19, May 2020, doi: 10.1016/j.clinimag.2020.01.008.
- [7] R. W. Filice et al., "Crowdsourcing pneumothorax annotations using machine learning annotations on the NIH chest X-ray dataset," J. Digit. Imaging, Nov. 2019, doi: 10.1007/s10278-019-00299-9.
- [8] M. Annarumma, S. J. Withey, R. J. Bakewell, E. Pesce, V. Goh, and G. Montana, "Automated Triaging of Adult Chest Radiographs with Deep Artificial Neural Networks," Radiology, vol. 291, no. 1, pp. 196–202, Jan. 2019, doi: 10.1148/radiol.2018180921



#### **CONCLUSION AND FUTURE WORK**



#### **Conclusion:**

It focused how image from given dataset (trained dataset) and past data set used to predict the pattern of pneumothorax disease using CNN model. This brings some of the following insights about pneumothorax disease prediction. The major benefit of the CNN classification framework is the ability to classify images automatically. The pneumothorax diseases mainly contribute to misshape and often can't be remedied because the patients are diagnosed too late with the diseases. We observed that deep learning is capable of achieving a relatively high diagnosis accuracy, is very much in agreement with human diagnostic performance, but still inferior to experienced radiologists in difficult-to-analyze cases.

In this study, we have discussed the overview of methodologies for detecting the abnormalities in pneumothorax images which includes collection of pneumothorax image data set, preprocessing techniques, feature extraction techniques.

#### **Future Work:**

Medical department wants to automate the detecting of pneumothorax disease from eligibility process (real time).

To optimize the work to implement in Artificial Intelligence environment.

