

## Pediatric Pneumonia Detection using Deep CNNs

Project Link: [Kaggle Notebook](#)

Dataset: [Chest X-Ray Images \(Pneumonia\)](#)

### 1. Introduction & Dataset Overview

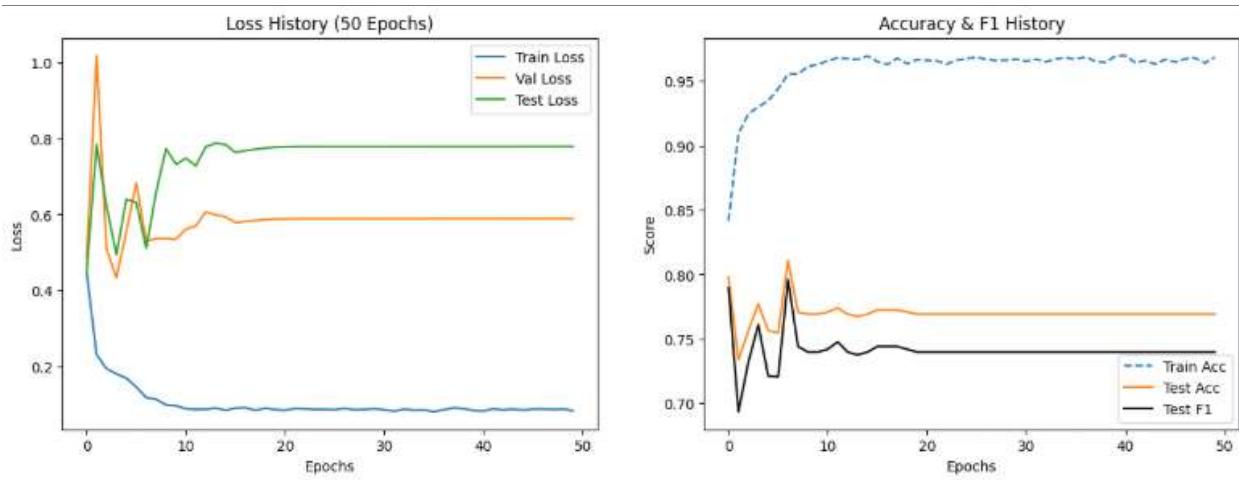
This study utilizes a retrospective cohort of pediatric chest X-ray images from patients aged one to five years at Guangzhou Women and Children's Medical Center. The primary objective is to develop a Convolutional Neural Network (CNN) capable of distinguishing between **Normal** lungs and those affected by **Pneumonia**.

- **Total Images:** 5,863 JPEG images.
- **Data Quality:** All scans underwent initial screening for quality control.
- **Ground Truth:** Diagnoses were graded by two expert physicians, with a third expert reviewing the evaluation set to eliminate grading errors.
- **Structure:** Data is partitioned into three directories: train, test, and val.

### 2. Methodology

The model was built using the **PyTorch Sequential API** with a three-block convolutional architecture. To ensure stability and prevent overfitting during the 50-epoch training phase, the following techniques were applied:

- **Feature Extraction:** Three Convolutional layers with increasing filter depth (32 to 64 to 128).
- **Regularization:** Dropout layers (0.2 to 0.5) to prevent co-adaptation of neurons.
- **Optimization:** Adam Optimizer with a starting Learning Rate of 0.001.
- **Learning Dynamics:** A ReduceLROnPlateau scheduler was implemented to lower the learning rate when validation loss stagnated.



### 3. Performance Evaluation

#### 3.1 Training Metrics Summary

Over 50 epochs, the model stabilized significantly. While the training loss showed a consistent downward trend, the validation and test accuracies reached a plateau, suggesting the model reached its maximum capacity for this specific architecture.

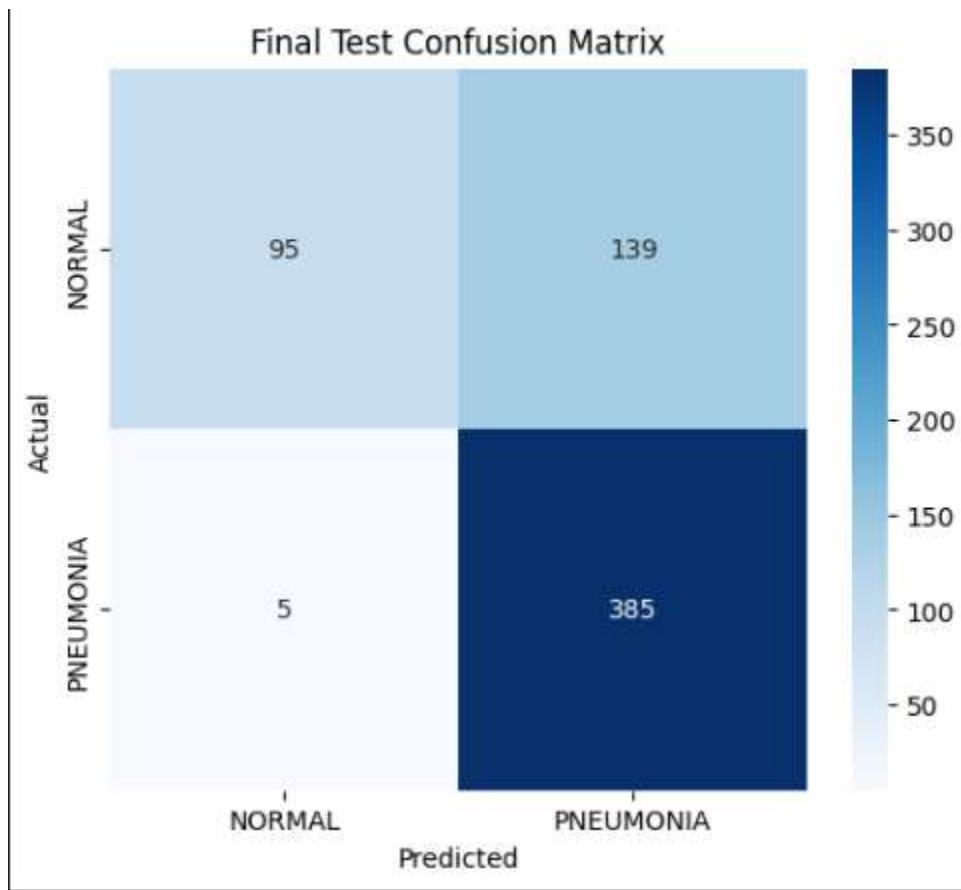
| Metric               | Start (Epoch 1) | End (Epoch 50) |
|----------------------|-----------------|----------------|
| <b>Train Loss</b>    | 0.4477          | 0.0834         |
| <b>Test Accuracy</b> | 79.81%          | 76.92%         |

#### 3.2 Classification Report (Test Set)

The model achieved a weighted average accuracy of **77%**.

| Class            | Precision | Recall | F1-Score | Support |
|------------------|-----------|--------|----------|---------|
| <b>NORMAL</b>    | 0.95      | 0.41   | 0.57     | 234     |
| <b>PNEUMONIA</b> | 0.73      | 0.99   | 0.84     | 390     |

| Class        | Precision | Recall | F1-Score | Support |
|--------------|-----------|--------|----------|---------|
| Weighted Avg | 0.82      | 0.77   | 0.74     | 624     |



#### 4. Discussion & Key Findings

- **Exceptional Sensitivity (Recall):** The model achieved a **99% Recall** for Pneumonia. In a medical context, this is critical; it means the AI almost never misses a sick patient (near-zero False Negatives).
- **Specificity Challenges:** The lower recall for "Normal" cases (41%) indicates that the model is "playing it safe" by over-identifying potential pneumonia patterns. This is likely due to the inherent class imbalance in the training data.

- **Convergence:** The learning rate scheduler effectively managed the plateauing loss in later epochs, keeping the test accuracy stable around 77%.

## 5. Conclusion

The developed CNN demonstrates high reliability in detecting positive cases of pneumonia, making it a viable screening tool to assist radiologists. Future work will focus on improving the precision for the Normal class through class-weight balancing and deeper architectures like ResNet.

## 6. Appendix:

### Sample Predictions:

