

Chest X-Ray Classification using Deep Learning

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Pneumonia Detection Challenge

Pneumonia, a **lung infection**, is a leading cause of death globally.

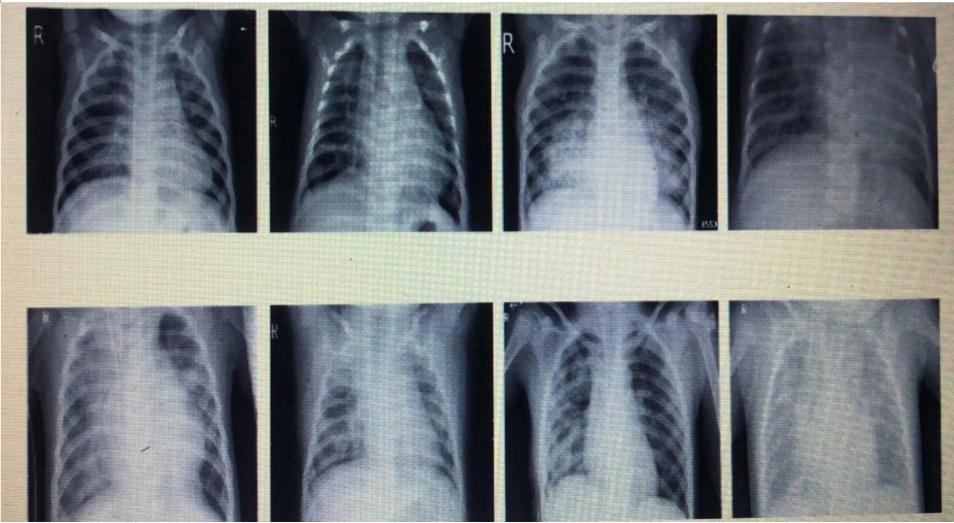
Diagnosis using chest X-rays is **labor-intensive** and prone to **human error.**

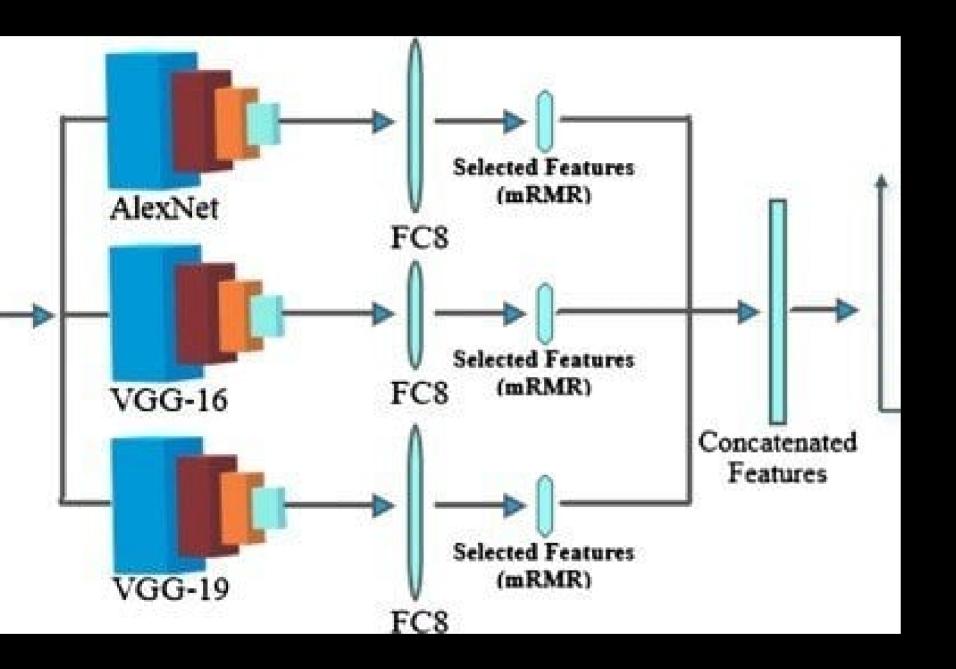
Early and accurate detection is crucial but challenging due to the subtlety of **visual cues** in X-rays.



Normal

PNEUMONIC





Our Solution

Automated Pneumonia Detection with Deep Learning

- A convolutional neural network (CNN) trained on a labeled chest X-ray dataset.
- Our model automates detection, reducing diagnosis time and improving accuracy.
- Deployed as an interactive web app using Hugging Face Spaces, accessible to healthcare professionals.

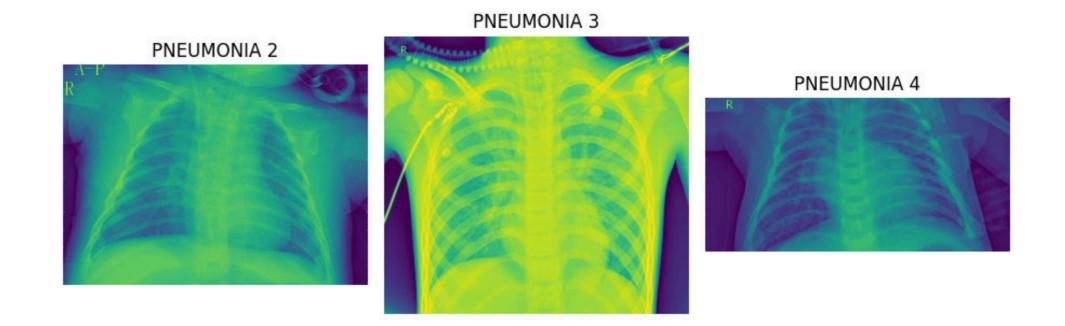
Dataset Overview

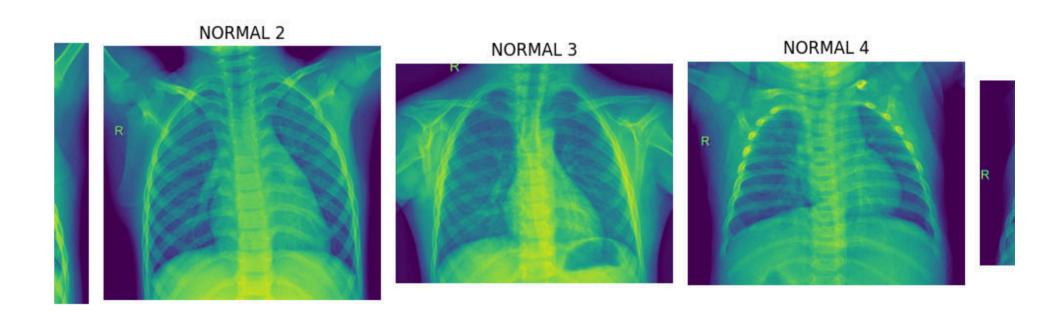
Train set:

• 1341 Normal, 3875 Pneumonia

Test set:

1. 234 Normal, 390 Pneumonia





input_layer (InputLayer) Output shape: (None, 224, 224, 3) conv2d (Conv2D) Input shape: (None, 224, 224, 3) Output shape: (None, 222, 222, 16) batch_normalization (BatchNormalization) Input shape: (None, 222, 222, 16) Output shape: (None, 222, 222, 16) activation (Activation) Output shape: (None, 222, 222, 16) Input shape: (None, 222, 222, 16) max_pooling2d (MaxPooling2D)

Model Architecture

GlobalAveragePooling2D
Dense layers with ReLU activation
Dropout for regularization
Final layer with sigmoid activation
for binary classification

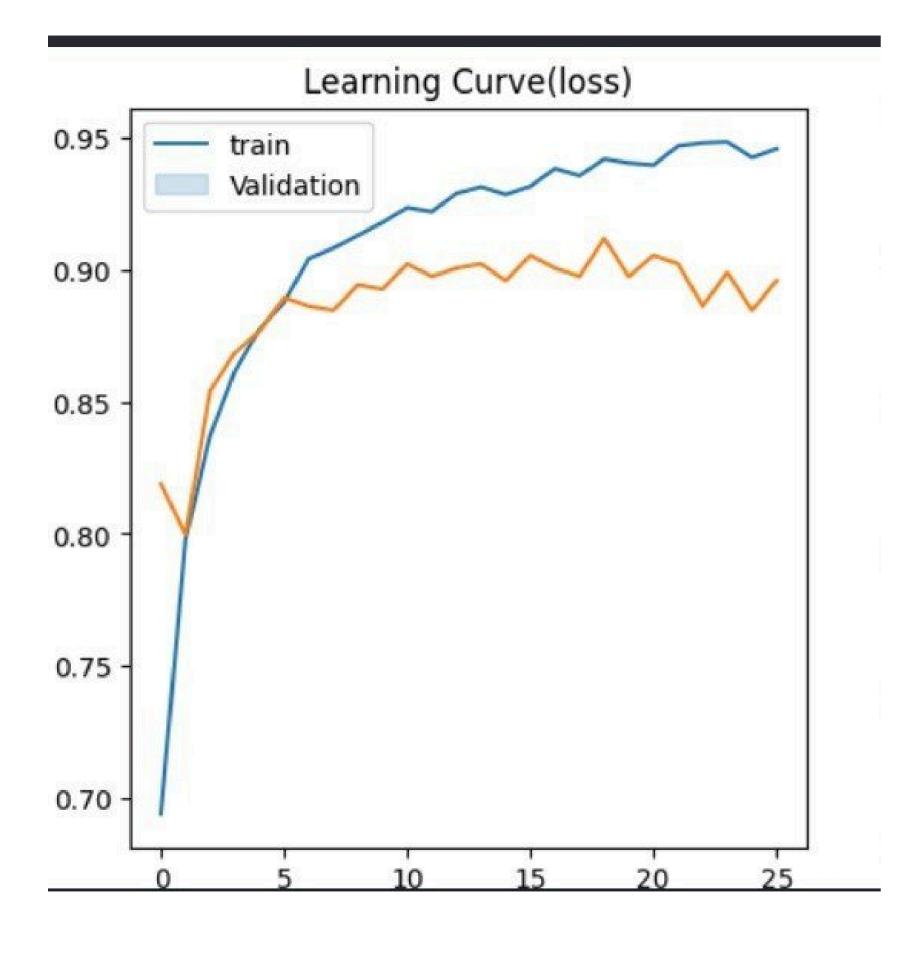
Training Process

Hyperparameters:

Optimizer: Adam

Loss: Binary Cross-Entropy

Learning rate: 5×10^-5



Model Evaluation

Evaluation metrics used:

Binary Accuracy, Confusion Matrix.

Accuracy on the validation set: 94%

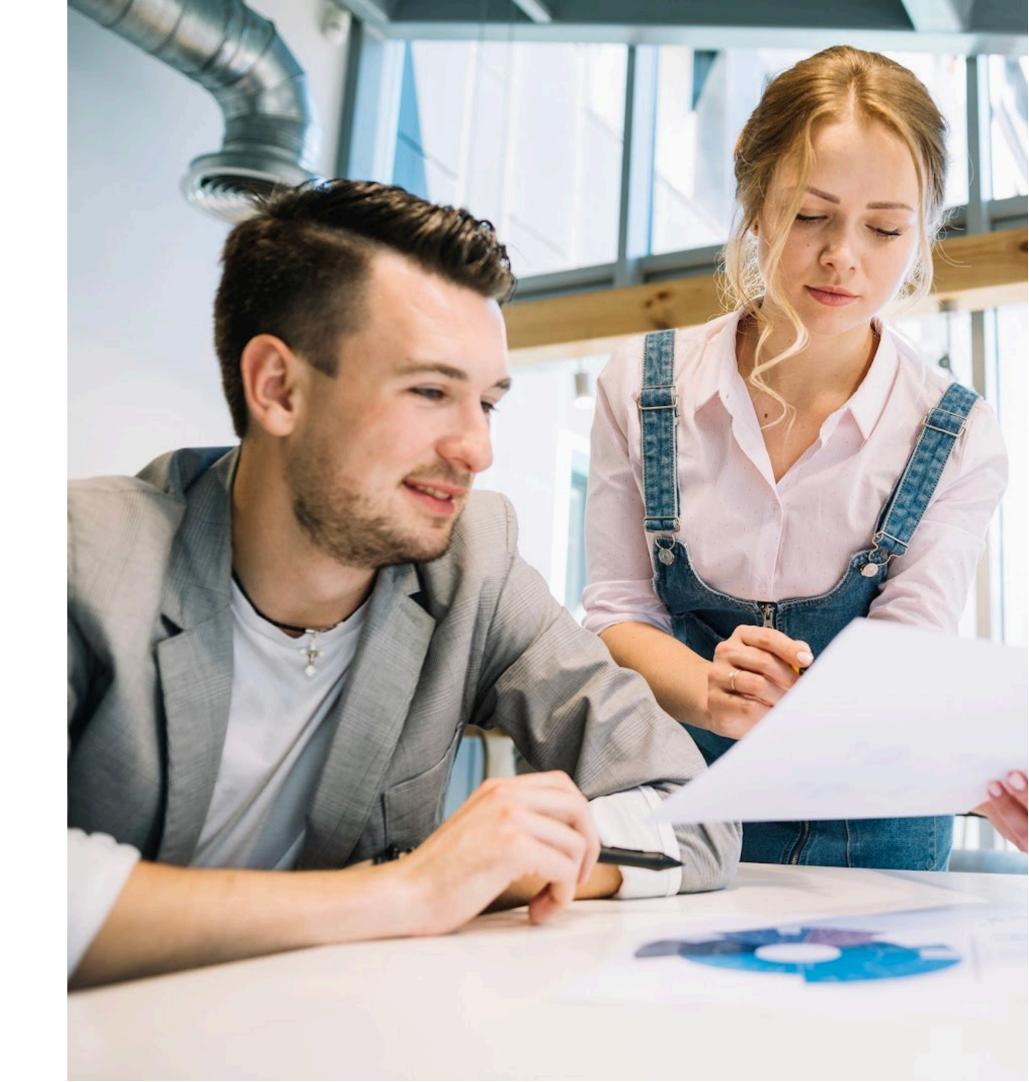
Overfitting: 4%

Loss decreased over epochs, improving performance.

Model Deployment

Save the trained model. Upload to Hugging Face repository.

Set up an interface using Gradio for real-time image classification.



Code Implementation

Model creation

(Pre_trained_model())Training code

block (fit method with callbacks) Model saving (save method for deployment)

```
def get model():
   #Input shape = [width, height, color channels]
   inputs = layers.Input(shape=(IMG_SIZE, IMG_SIZE, 3))
   # Block One
  x = layers.Conv2D(filters=16, kernel_size=3, padding='valid')(inputs)
  x = layers.BatchNormalization()(x)
  x = layers.Activation('relu')(x)
  x = layers.MaxPool2D()(x)
  x = layers.Dropout(0.2)(x)
   # Block Two
   x = layers.Conv2D(filters=32, kernel_size=3, padding='valid')(x)
  x = layers.BatchNormalization()(x)
   x = layers.Activation('relu')(x)
  x = layers.MaxPool2D()(x)
  x = layers.Dropout(0.2)(x)
   # Block Three
  x = layers.Conv2D(filters=64, kernel size=3, padding='valid')(x)
  x = layers.Conv2D(filters=64, kernel_size=3, padding='valid')(x)
  x = layers.BatchNormalization()(x)
  x = layers.Activation('relu')(x)
  x = layers.MaxPool2D()(x)
  x = layers.Dropout(0.4)(x)
   # Head
   #x = layers.BatchNormalization()(x)
  x = layers.Flatten()(x)
  x = layers.Dense(64, activation='relu')(x)
  x = layers.Dropout(0.5)(x)
   #Final Layer (Output)
  output = layers.Dense(1, activation='sigmoid')(x)
   model = Model(inputs=[inputs], outputs=output)
```

```
##2
```

```
model pretrained = Pre trained model()
model pretrained.compile(loss='binary crossentropy'
              , optimizer =optimizers.Adam(learning rate=5e-5), metrics=['binar'
model pretrained.summary()
##1 perfect
history pretrained=model pretrained.fit(
    train_generator,
   epochs=20,
    validation data=val generator,
    callbacks=[model_checkpoint_callback,Early_Stopping,reduce_lr]
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    train generator,
    epochs=20,
    validation data=val generator,
    callbacks=[model checkpoint callback, Early Stopping, reduce lr]
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Results & Conclusion

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