# **Model Development Phase**

**Project Name:** COVID-19 Chest X-Ray Image Classification

### **Model Selection Report**

In this project, multiple deep learning approaches were evaluated for classifying chest X-ray images into three categories: COVID-19, Bacterial Pneumonia, and Normal.

Factors such as training accuracy, validation accuracy, confusion matrix, and loss curves were considered to select the most suitable model.

### Models:

## Model 1: VGG16 (Transfer Learning):

- Pretrained CNN (VGG16) using ImageNet weights.
- Added custom dense layers (Flatten → Dense → Dropout → Softmax).
- Achieved strong performance and generalization due to transfer learning.
- Selected as the final model.

### **Model Development Steps:**

Model steps	Description
Step 1: Dataset Loading & Preprocessing	Loaded chest X-ray dataset from train, validation, and test folders. Resized all

images to 64×64, normalized pixel values (rescale 0–1), and generated batches with ImageDataGenerator.

```
[1]: import os
    os.listdir("chest_xray_covid/Data")
[1]: ['test', 'train']
[2]: Image_size = [64, 64]
    train_path = 'chest_xray_covid/Data/train'
    test_path = 'chest_xray_covid/Data/test'
```

## Step 2: Splitting Dataset

Used ImageDataGenerator with Validation\_split = 0.2 to automatically split training data into 80% training and 20% validation. Separate test set used for final evaluation.

```
[3]: import tensorflow as tf
     from tensorflow import keras
    from tensorflow.keras.preprocessing.image import ImageDataGenerator
    # Rescale images
train_datagen = ImageDataGenerator(
        validation_split=0.2 # <-- split 20% of train into validation
    test datagen = ImageDataGenerator(rescale=1./255)
    # Training set (80% of train data)
    training_set = train_datagen.flow_from_directory(
        'chest_xray_covid/Data/train',
target_size=(64, 64),
        batch_size=32,
        class_mode='categorical',
        subset='training', # <-- important</pre>
        seed=42
   # Validation set (20% of train data)
   val_set = train_datagen.flow_from_directory(
        'chest_xray_covid/Data/train',
       target_size=(64, 64),
       batch size=32.
       class_mode='categorical',
        subset='validation', # <-- important</pre>
       seed=42
   # Test set (uses separate test folder)
   test_set = test_datagen.flow_from_directory(
        'chest_xray_covid/Data/test',
        target_size=(64, 64),
       batch_size=32,
        class_mode='categorical'
```

## Step 3: Label Encoding

Labels (COVID-19, Normal, Bacteria) were automatically one-hot encoded by flow\_from\_directory.

```
# Training set (80% of train data)
                                                                               training_set = train_datagen.flow_from_directory(
                                                                                    'chest_xray_covid/Data/train',
                                                                                   target_size=(64, 64),
                                                                                   batch_size=32,
                                                                                   class_mode='categorical',
                                                                                   subset='training', # <-- important</pre>
                                                                               # Validation set (20% of train data)
                                                                               val_set = train_datagen.flow_from_directory(
                                                                                    'chest_xray_covid/Data/train',
                                                                                   target_size=(64, 64),
                                                                                   batch_size=32,
                                                                                   class_mode='categorical',
                                                                                   subset='validation', # <-- important</pre>
                                                                                   seed=42
                                                                               # Test set (uses separate test folder)
                                                                               test_set = test_datagen.flow_from_directory(
                                                                                    'chest_xray_covid/Data/test',
                                                                                   target_size=(64, 64),
                                                                                   batch_size=32,
                                                                                   class_mode='categorical'
Step 4: Training the model
                                                                            Custom CNN trained as baseline.
                                                                            VGG16 Transfer Learning fine-tuned with
                                                                            additional dense layers for classification.
                                                                               from keras.applications import VGG16
                                                                                from tensorflow.keras.models import Model
from tensorflow.keras.layers import Dense, Flatten, Dropout, Input
                                                                            5]: base_model = VGG16(input_shape = (64, 64, 3), include_top=False , weights="imagenet")
                                                                               x = base_model.output
x = Flatten()(x)
x = Dense(1024, activation = 'relu')(x)
x = Dropout(0.5)(x)
                                                                               output = Dense(3, activation = 'softmax')(x)
model = Model(inputs = inp, outputs = output)
                                                                            71: model.summarv()
Step 5: Evaluation
                                                                            Compared models using
                                                                            training/validation accuracy, loss curves,
                                                                            and confusion matrix.
                                                                            VGG16 achieved higher accuracy and
                                                                            stability, making it the chosen model.
                                                                             ]: test_loss, test_acc = model.evaluate(test_set)
                                                                                             - 36s 851ms/step - accuracy: 0.8968 - loss: 0.4168
                                                                             ]: history = model.fit(training_set, epochs = 2, validation_data = val_set)
                                                                                             —— 611s 5s/step - accuracy: 0.6361 - loss: 1.0127 - val_accuracy: 0.6663 - val_loss: 0.7671
                                                                                              — 597s 5s/step - accuracy: 0.7711 - loss: 0.5901 - val_accuracy: 0.8638 - val_loss: 0.4536
Step 6: Prediction
                                                                            The final VGG16-based model was used
                                                                            to predict disease class on unseen test
                                                                            X-ray images.
```

```
o, ........
16]: from flask import Flask, request, render_template_string
       from werkzeug.utils import secure_filenam
      import numpy as np
from tensorflow.keras.preprocessing import image
from tensorflow.keras.models import load_model
       import os
       # Initialize Flask app
       app = Flask(__name__)
       UPLOAD_FOLDER = "uploads"
      os.makedirs(UPLOAD_FOLDER, exist_ok=True)
app.config['UPLOAD_FOLDER'] = UPLOAD_FOLDER
       # Model and class labels
       class_labels = ["COVID-19", "Normal", "Bacteria"]
      try:
    model = load_model("covid.h5") # replace with your model path
      except Exception as e:
    print(f"Error loading model: {e}")
            model = None
 IMG_SIZE = (64, 64) # Change to match your model input size
  # Home route
 @app.route('/')
 def home():
       return render_template_string("""
           <h2>Chest X-ray Classification</h2>
<form action="/predict" method="post" enctype="multipart/form-data">
                 <label>Select an X-ray image:</label>
<input type="file" name="file" required>
<button type="submit">Predict</button>
            </form>
 # Prediction route
  @app.route('/predict', methods=['POST'])
 def predict():
     if model is None:
            return "Model is not loaded. Please check server logs."
       if 'file' not in request.files:
            return "No file uploaded."
       # Save the uploaded file filepath = os.path.join(app.config['UPLOAD_FOLDER'], secure_filename(file.filename)) file.save(filepath)
           # Preprocess the image
img = image.load_img(filepath, target_size=IMG_SIZE)
           img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0) / 255.0 # normalize
            # Predict
            preds = model.predict(img_array)
           predicted_class_idx = np.argmax(preds, axis=1)[0]
confidence = np.max(preds)
predicted_label = class_labels[predicted_class_idx]
       except Exception as e:
    return f"Prediction failed: {e}"
       # Return the result
       return render_template_string("""
```