

## Data Collection and Preprocessing

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

The dataset of chest X-ray images was prepared for deep learning classification to detect COVID-19, Bacterial Pneumonia, and Normal cases using a Convolutional Neural Network (CNN) with transfer learning (VGG16)

### PREPROCESSING STEPS :

SECTION	DESCRIPTION
Data Overview	Collected chest X-ray dataset organized into train and test folders, each containing subfolders for classes (Normal, COVID-19, Bacteria).
Image Resizing	Resized all images to 64×64 pixels for uniform input into the CNN model.
Rescaling	Applied pixel normalization by rescaling image values from 0–255 → 0–1.
Data Augmentation	Used ImageDataGenerator to apply transformations and split 20% of training data into validation automatically.
Train / Validation split	Training set = 80% of train folder, Validation set = 20% of train folder (via subset in ImageDataGenerator).
Test set	Loaded separately from the test directory without augmentation, only rescaling.
Label Encoding	Automatically handled by flow_from_directory, mapping class subfolders to categorical one-hot labels.
Model Preparation	Used VGG16 pretrained on ImageNet as

	the base model (without top layers). Added custom layers: Flatten → Dense → Dropout → Output.
Visualization	Generated plots comparing training accuracy/loss vs validation accuracy/loss across epochs.

## Data Preprocessing Code Snapshots :

SECTION	CODE
Data Overview	<pre>[1]: import os       os.listdir("chest_xray_covid/Data")  [1]: ['test', 'train']</pre>
Image Resizing	<pre>[2]: Image_size = [64, 64]        train_path = 'chest_xray_covid/Data/train'       test_path = 'chest_xray_covid/Data/test'</pre>
Train / Validation split	<pre>: import tensorflow as tf   from tensorflow import keras   from tensorflow.keras.preprocessing.image import ImageDataGenerator    # Rescale images   train_datagen = ImageDataGenerator(       rescale=1./255,       validation_split=0.2 # &lt;-- 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', # &lt;-- important       seed=42   )</pre>

	<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',  # &lt;-- important     seed=42 ) </pre>
Test set	<pre> # 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' ) </pre>
Model Preparation	<pre> from keras.applications import VGG16 from tensorflow.keras.models import Model from tensorflow.keras.layers import Dense, Flatten, Dropout, Input  base_model = VGG16(input_shape = (64, 64, 3), include_top=False, weights="imagenet")  inp = base_model.input 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)  model.summary()  model.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])  test_loss, test_acc = model.evaluate(test_set)  C:\Users\lenovo\AppData\Local\Programs\Python\Python313\Lib\site-packages\keras\sources\trainers\data_adapters\py_dataset_adapter.py:10: PyDataset class should call 'super().__init__()' in its constructor. '**kwargs' can include 'workers', 'use_multi_workers', 'use_multi_workers' or 'use_multi_workers' as they will be ignored. self._warn_if_super_not_called() 41/41 ----- 71s 2s/step - accuracy: 0.0988 - loss: 1.5700  history = model.fit(training_set, epochs = 5, validation_data = val_set) </pre>
Visualization	<pre> import matplotlib.pyplot as plt  # Assuming you already trained your model # history = model.fit(...)  # Plot accuracy plt.figure(figsize=(8, 6)) plt.plot(history.history['accuracy'], label='Training Accuracy') plt.plot(history.history['val_accuracy'], label='Validation Accuracy') plt.title('Model Accuracy') plt.xlabel('Epoch') plt.ylabel('Accuracy') plt.legend(loc='lower right') plt.show()  # Plot Loss plt.figure(figsize=(8, 6)) plt.plot(history.history['loss'], label='Training Loss') plt.plot(history.history['val_loss'], label='Validation Loss') plt.title('Model Loss') plt.xlabel('Epoch') plt.ylabel('Loss') plt.legend(loc='upper right') plt.show() </pre>