SYSTEM DESIGN

SYSTEM DESIGN AND METHODOLOGY

In the automation of multiple processes, machine learning plays a critical role. The proposed architecture was designed with that goal in mind, and it is based on machine learning methodologies. Especially in the case of detecting and categorizing images into various disease categories. This section has been structured such that the topic begins with the device specifications and data acquisition for the data used in the currently suggested approach using Keras and TensorFlow. The second point of discussion would be CNN (Convolution Neural Network). Image augmentation and image pre-processing would be the next thing to focus on. The fourth point of discussion would be Adam (Optimization algorithm).

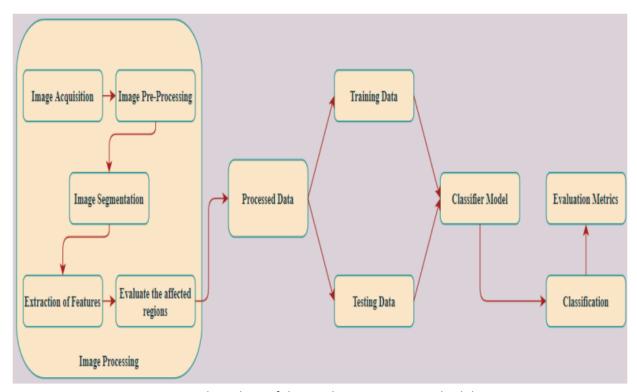


Figure 1. Flow Chart of the Machine Learning Methodology

1) Keras

Keras is a high-level, deep learning API developed by Google for implementing neural networks. It is written in Python and is used to make the implementation of neural networks easy. It also supports multiple backend neural network computation. Keras has features such as:

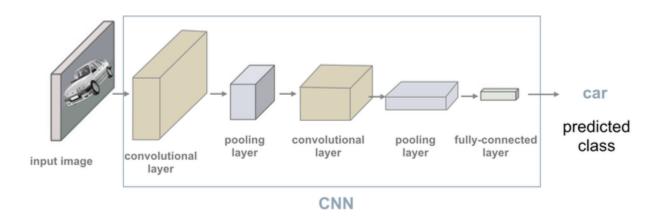
a) It runs smoothly on both CPU and GPU. b) It supports almost all neural network models.

2) TensorFlow

TensorFlow is an open-source library developed by Google primarily for deep learning applications. It also supports traditional machine learning. TensorFlow was originally developed for large numerical computations without keeping deep learning in mind. However, it proved to be very useful for deep learning development as well. TensorFlow accepts data in the form of multi-dimensional arrays of higher dimensions called tensors. Multi-dimensional arrays are very handy in handling large amounts of data.

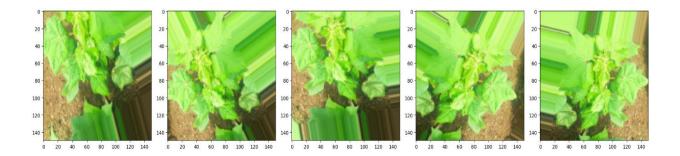
3) CNN (Convolution Neural Network)

A Convolutional Neural Network (CNN) is a type of Deep Learning architecture commonly used for image classification and recognition tasks. Convolutional neural networks are distinguished from other neural networks by their superior performance with image, speech, or audio signal inputs. They have three main types of layers, which are: 1) Convolutional layer 2) Pooling layer 3) Fully-connected (FC) layer. With each layer, the CNN increases in its complexity, identifying greater portions of the image. Earlier layers focus on simple features, such as colors and edges. As the image data progresses through the layers of the CNN, it starts to recognize larger elements or shapes of the object until it finally identifies the intended object.



4) Image Augmentation

Image augmentation is a technique of applying different transformations to original images which results in multiple transformed copies of the same image. Each copy, however, is different from the other in certain aspects depending on the augmentation techniques you apply like shifting, rotating, flipping, etc. These image augmentation techniques not only expand the size of your dataset but also incorporate a level of variation in the dataset which allows your model to generalize better on unseen data. Also, the model becomes more robust when it is trained on new, slightly altered images.



5) Image Pre-Processing

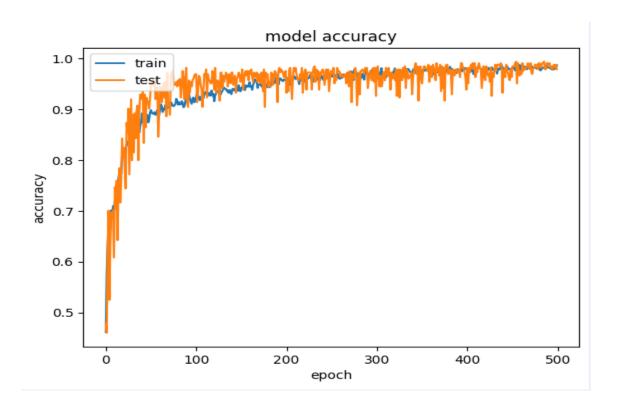
The use of computer algorithms to perform image processing on digital images is known as image pre-processing. We can detect the plant by analyzing the image with a specific algorithm. We use a similar approach for image processing and detection with a specific algorithm. The image quality is critical in this process; we can't use the algorithm if the image isn't clear.

6) Adam (optimization algorithm)

Adam is an optimization algorithm that can be used instead of the classical stochastic gradient descent procedure to update network weights iterative based in training data. The algorithm is called Adam. It is not an acronym and is not written as "ADAM". *The name Adam is derived from Adaptive Moment Estimation*. Some of the attractive benefits of using Adam are, straightforward to implement, computationally efficient, little memory requirements.

ACCURACY

```
■ Crop_Disease_Predictor_Model_Training.ipynb ×
                                                                                                          $ ▷ □
+ Code + Markdown | ▶ Run All ■ Clear All Outputs | ■ Outline ···
                                                                                                          ■ Select K
   Output exceeds the size limit. Open the full output data in a text editor
    Epoch 1: val_accuracy improved from -inf to 0.47840, saving model to D:\Plant Disease Predictor\Project\Project\Trained Model\cotton_plant_dise
   61/61 [====
                             ====] - 47s 762ms/step - loss: 1.2330 - accuracy: 0.4613 - val_loss: 1.0473 - val_accuracy: 0.4784
   Epoch 2/500
                              ====] - ETA: 0s - loss: 1.0269 - accuracy: 0.5761
   61/61 [===
   Epoch 2: val_accuracy did not improve from 0.47840
                              ====] - 33s 540ms/step - loss: 1.0269 - accuracy: 0.5761 - val loss: 1.0837 - val accuracy: 0.4599
   61/61 [====
   Epoch 3: val_accuracy improved from 0.47840 to 0.61728, saving model to D:\Plant Disease Predictor\Project\Project\Trained Model\cotton_plant_
                          =======] - 33s 545ms/step - loss: 0.9124 - accuracy: 0.6325 - val_loss: 0.8947 - val_accuracy: 0.6173
   Epoch 4/500
   61/61 [====
                            =====] - ETA: 0s - loss: 0.8270 - accuracy: 0.6709
   Epoch 5: val_accuracy did not improve from 0.70062
                              ====] - 33s 541ms/step - loss: 0.8272 - accuracy: 0.6689 - val_loss: 1.1228 - val_accuracy: 0.5247
   Epoch 6/500
   61/61 [=====
   Epoch 7/500
   Epoch 500/500
   61/61 [=====] - ETA: 0s - lo
Epoch 500: val_accuracy did not improve from 0.99383
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OUTPUT

