

Detection of Plant Leaf Disease Using Image Processing

Dr. P. Prabhakaran¹, P. Raguraman², Dr. R. Saran Kumar³, V. Vaneesha⁴, K. Venkata Ratnam⁵

^{1, 2, 4, 5} Department of Computer Science and Engineering,

³ Department of Electronics and Communications Engineering

^{1,2,3,4} QIS College of Engineering and Technology, Ongole, Andhra Pradesh, India

⁵ Engineering and Technology Program, GVPCDPGC(A)

¹prabhakaran.p@qiscet.edu.in, ²raguraman.p@qiscet.edu.in, ³saran.r@qiscet.edu.in

vaneesha.v@qiscet.edu.in, ⁵kvenkataratnam@gvpcdpgc.edu.in

Corresponding Author Mail: qispublications@qiscet.edu.in

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Abstract

There is critical efficiency and monetary harm because of plant infections and the general nature of ranch items are both reduced. Distinguishing plant illnesses has become more significant in the reconnaissance of tremendous fields of yields in the present day. With regards to sickness the board, ranchers struggle with changing starting with one methodology then onto the next. The ordinary strategy for distinguishing and recognizing plant illnesses is proficient unaided eye investigation. In this exploration, we look at the need for a basic method for distinguishing plant leaf illness that would help horticultural developments. Compelling observing strategies might be worked with by early information on crop wellbeing and sickness discovery. Crop yields will ascend because of this technique. Moreover, the benefits and downsides of every one of these imminent methodologies are talked about in this review. Picture catch, picture investigation, extraction of elements, and arrangement in light of brain networks are all important for the cycle.

Introduction:

The majority of the country's people are farmers, making it a country that depends heavily on agriculture. Agricultural research seeks to increase output and food quality while minimizing expenses and maximizing profit. Agrochemicals, seeds, and soil combine to create agricultural goods. The most crucial items in agriculture are fruits and vegetables. If you want to generate more valuable things, you must practice product quality control. Numerous studies have found that plant diseases might potentially reduce the quality of agricultural products. Diseases are defects in a plant's normal processes, such as photosynthesis, transpiration, fertilization, pollination, and germination. These illnesses are caused by pathogens like fungi, bacteria, and viruses as well as unfavorable environmental conditions. Therefore, it is crucial to detect plant diseases early on [1]. Professional surveillance may be overly expensive and time-consuming for farmers. Therefore, it is crucial to look for a quick, low-cost, and accurate method of identifying illness based on symptoms that appear on plant leaves. Machine vision may now be used to automate inspection, process management, and robot navigation. The purpose of this study is to investigate potential methods for diagnosing plant disease based on leaf texture. The leaf offers many advantages over flowers and fruits throughout the year. An efficient, automated, and accurate disease detection system is required. This review paper presents an overview of various feature extraction and classification approaches for the identification and categorization of plant diseases.



Figure 1. Infected leaf of cotton plant

Literature Survey :

There are numerous trainings that outline how to identify and address issues. The focus of this discourse is on techniques that show how to put them into practice and increase the likelihood of discovering damaged cotton leaves.

Mr. V. A. Gulhane and Dr. A. A. Gurjar made the diagnosis [1].

A method that allows you to save colour and identify appearances backwards self-organizing piece map of a system of neurons broadcast finding a sick component on a plant's greenery and Plant leaf viruses are arranged according to texture behavior.

Arivazhagan et al. [2] developed an algorithm to process the data and a technique for converting the supplied colour Green pixels are hidden after forming to create the red, green, and blue mechanisms of an RGB image. The data is used to extract figure and form texturing features. The Minimum Space is utilised in the track to order the objects. for standard evaluation and assistance, vector machines (SVMs).

They have created a framework they name A Framework with Dheeb Al Bashish [3]. [4] Leaf and Branch Plant Viruses: Documentation and Classification It divides and transforms utilizing the K-Means algorithm.

RGB images into HIS Tonal scheme. Color and surfacefeatures are then considered.

A classifier that employs a neural network with statistics. The use of classification is for categorization.

Radial potential has been hypothesized by ElhamOmrani, et al[4]. functional support as a foundation Apple's vector regressions pesticides have been used to identify leaf diseases, The three-part procedure is as follows:

The leaf images were initially stored in the device-independent colour format RGB. The photographs have to be scaled and transformed to the hardware-independent CIELAB colour space in order to be separated.

The second method is region-based extraction of the polluted area. Segmentation was used in this. When using K-means clustering, it's crucial to a technique applied in the segmentation phase.

Characteristics of the twelfth extract based on colour, shape, and texture. In actuality, frequently applied to describe .As a result, the features, wavelet, and grey level co-occurrence matrix It was necessary to use particular strategies. Incorporating the support vector regression (SVR) method to classify apples based on leaf diseasesRiceailment has been hypothesised by SantanuPhadikar, et al[6].

A software prototype method for identifying rice illness based on photographs of different rice plants that contain the disease is as explained in Identification Using Pattern Recognition Methods. Digital cameras are used to take pictures of injured rice plants, and image growth and image segmentation algorithms are used to identify ill areas of the plants. An ANN is then used to categorize the damaged leaf flesh.

Edge, CYMK colour feature, GA feature, Color, and Texture variations features are extracted using the enhanced PSO feature selection strategy in this study. P. Revathi et al. [7] described this technique for identifying cotton leaf spot infections. Utilizing SVM, BPN, and Fuzzy with Edge Selection & Classification, the collected features were evaluated.

Y.Sanjana, AshwathSivasamy, and et al.... [8]

When the images are submitted to the remote server, the images are analyzed and presented to a panel of experts for their evaluation. Spoiled parts in a photograph can be recognized and categorized using computer vision techniques. The disease-affected lesions are divided into groups using a simple colour difference method. The expert can review the study's results and provide the farmers with feedback through a mobile phone notification. This work aims to develop a crop disease detection picture recognition system.

The initial stage of image processing begins with the scanning of a colour image of a sick leaf. An approach called mathematical morphology is used to segment these images. The three groups of illnesses were then divided using these characteristics together with a classification strategy known as the membership function.

This includes BhumikaPrajapati, VipulDabhi, and others. [9] Using image processing and machine learning approaches, cotton leaf disease was identified and classified. Throughout the conversation, the findings of a poll on backdrop removal and segmentation strategies were also mentioned. During this research, a conversion from RGB to HSV colour space for background removal was found. Furthermore, we found that thresholding outperforms other background noise removal techniques in terms of quality of outcomes. After conducting colour processing, we used thresholding to the masked image to create a binary image..

segmentation using green background-removed image pixels as a mask. In order to extract characteristics specific to an illness, this is crucial. We found that SVM performed well in terms of accuracy for categorising illnesses. Three of our suggested procedures—image acquisition, image preprocessing, and image segmentation—have all been put into practise. Five main components make up our suggested work.

"P. Revathi, M. Hemalathy, and others..." [10] Using image edge recognition and segmentation techniques, the collected photos are first improved after being taken. The R, G, and B colour feature images are used to segment the photos (disease spots). Then, depending on these attributes, a pest

control strategy is put into place after disease areas are identified using an image feature extraction procedure.

In this work, illnesses are analysed and classified using cotton leaf spot, cotton leaf colour segmentation, and edge-based image segmentation.

He's SushilPatil and PramodLandge and et al. [11] Using Image Processing, We suggest and evaluate a fresh software technique for identifying and categorizing plant diseases. In rural India, there are few agricultural experts who can evaluate crop images and offer guidance. Due to the delay in expert responses, farmers frequently receive expert advice too late. To recognize agricultural diseases or other problems that could affect crops from images and give farmers the most up-to-date knowledge via SMS, this article will develop IP algorithms based on colour, surface, and shape. Through the more efficient use of chemicals, these technologies will increase productivity and improve output quality.

Methodology:

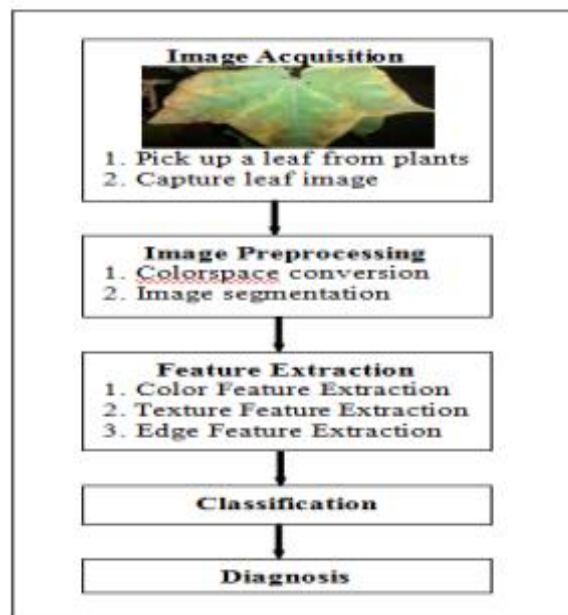


Figure 2. The Basic Steps for Disease Detection Algorithm

The gathering of images to begin, a digital camera with the requisite resolution is used to capture high-resolution RGB colour images. The project precisely specifies the structure of an image database. The picture database itself improves the efficiency of the classifier, which establishes the algorithm's durability. image processing and image segmentation

Unwanted distortions are controlled and background and noise from the image data are removed during pre-processing. Capabilities for image processing and analysis are improved.

The RGB colour image is transformed using the HIS and CIELAB colour spaces. Both HSI and CIELAB were based on human vision when they were created as colour independent space models.

The initial step in finding the polluted area is segmentation.

For segmentation, edge detection and k-means clustering are frequently employed.

After segmentation, the contaminated region is defined using a variety of factors. Color, texture, and form are frequently used to depict a place.

[9] Color attributes are crucial for establishing the visual environment, recognizing objects, and communicating information.

One of an object's most important identifying features is its texture. Its effective use as a strong regional descriptor aids with image retrieval.

Contrast, homogeneity, dissimilarity, energy, and entropy are used to characterize texture. One of the most fundamental characteristics that may be used to characterize an image's content is its shape.

Classification

No new diseases can be found at this time. It is establishing a rule by categorizing each ailment into one of the predetermined groups and based on specific traits [4]. It is common practice to use classifiers like the SVM and the ANN.

Conclusion:

In this work, image processing techniques that have been applied to identifying plant diseases in a variety of plant species are reviewed and summarized. The most popular techniques for identifying plant diseases include BPNN, SVM, K-means clustering, and SGDM. These techniques are used to analyze the leaves of both healthy and injured plants. Obstacles in these processes include, for instance, optimizing the strategy for a specific plant leaf disease and automating the technique for continuous automated monitoring of plant leaf diseases in real-world field situations. According to a study printed in the journal Plant Disease, this disease detection technique offers a lot of promise but also some limitations. The current research could be enhanced as a result.

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