



Detection of Powdery Mildew Disease of Beans in India : A Review

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ABSTRACT

Powdery mildew disease of beans in India causes major economic losses in agriculture. For sustainable agriculture, detection and identification of diseases in plants is very important. In this review, we are trying to identify the powdery mildew disease of beans crop by using some image processing and pattern recognition techniques and comparing with molecular and spectroscopic techniques. Early information on crop health and disease detection can facilitate the control of diseases through proper management strategies. The present review recognizes the need for developing a rapid, cost-effective, and reliable health monitoring techniques that would facilitate advancements in agriculture. These technologies include image processing and pattern recognition based plant disease detection methods

Keywords: beans, powdery mildew, detection, image processing, pattern recognition.

INTRODUCTION

Pulse production has been stagnant in Asian countries leading to steady decline in per capita availability over the past 50 years. In India 13.00 - 15.00 million metric tons (mmt) of peas and pulses produced which is about 25% of the global production. India is the largest producer as well as

major importer of peas/pulses in the World. Pulse is a common name for all varieties of dry peas, lentils, beans and chickpeas. Almost all varieties of pulses, except lupins, faba beans and dun peas, are grown in India.

Beans are intricately woven into the fabric of human history. The primary 'everlasting cultures'

advanced when hunter-gatherers and nomadic persons started tilling the earth, establishing programs of agriculture and beans were among the first cultivated crop. Among beans soybean (*Glycine max* (L.) Merr.) and peanut (*Arachis hypogea* L.) are the 3rd most essential food legume crop worldwide¹. Cultivated beans are consumed in the form of dry edible beans (French bean (Rajma), soybean, mung bean, lima bean and azuki bean) as the mature dry seed after rehydration, and snap beans (e.g., green, string, Haricot bean) as fleshy immature pods. Dry beans are further divided into distinct market classes based on seed characteristics and snap bean classes are based on pod characteristics and plant type².

Powdery Mildew Diseases Of Beans

Diseases are often encountered through the growing stage, the production stage which is most favorable stage for various plant pathogens like crop residue-borne fungi and other soil-surface-inhabiting microbes and postharvest storage^{3,4}. The most vital factor for the development of certain plant pathogens and plant diseases is that the presence of wetness (moisture) at the soil-air interface or within the top few millimeters of soil^{5,6}. Soil surface moisture, usually related to the precipitation events or dew formation in the canopy^{7,8}, is plausible to be one of the most important environmental factor affecting the growth of plant pathogens to cause many bacterial and fungal plants diseases².

Powdery mildew, a fungal disease of beans, is caused by many different species of fungi in the order Erysiphales and affects a wide range of beans plants. It is a foliar disease of beans, as its symptoms are quite distinctive in form of white powdery spots on the leaves and stems of the infected plants (Figure1). The most affected part in the infected plants is the lower surface of leaves, but the mildew can appear on any above-ground part of the plant. Disease progresses in the form of white spots of asexual spores that get larger and denser and may spread up and down the stem of the plant¹³.

Here, in this review, we are focusing on powdery mildew disease i.e. foliar disease of beans and most critical leaf surface conditions; favorable temperature (22-27 °C) and leaf surface humidity

(High relative humidity at night, Low relative humidity during day) are required for occurrence of powdery mildew disease⁹. These favorable conditions cause greater the risk of infection and the greater the number of infections per leaf. Even powdery mildew diseases have been reported to be more severe with prolonged periods of these critical conditions^{10,12}. Thus, disease management prediction requires leaf wetness assessment.

Management Of Powdery Mildew Disease

In agriculture, there are various methods that can be applied for controlling powdery mildew disease i.e. fertilizers, chemical methods, genetic resistance, and careful farming methods. It is important to be aware of powdery mildew and its management as the resulting disease can significantly reduce crop yields¹⁴.

Carefully timing application and rate of nitrogen as fertilizer (less than 70 pounds/acre) can be used to eliminate the progress of disease by altering planting density. Another ways to manage disease are crop rotation, eliminating the presence of volunteer in agriculture field, use of highly resistance varieties¹⁵.

Chemical control is possible with fungicides such as triadimefon and propiconazole. Bennet *et al.* (1984) used calcium silicate slag for treatment of powdery mildew disease of wheat that helps the plant cells to defend against fungal attack by degrading haustoria and by producing callose and papilla¹⁶. Some studies have shown milk's effectiveness as comparable to some conventional fungicides¹⁷, and as better as benomyl and fenarimol at higher concentrations¹⁸. Some home gardeners and small scale farmers have been used milk for the treatment of mildew disease, by diluting with water and sprayed on infected plants at the initial stage, or as a preventative measure, with repeated weekly application on the plants.

Some agricultural universities and institutes have been releasing resistance varieties of many crops that can be an economically and efficient method in disease reduction. The most common and serious disease-resistant varieties should be selected on the basis of local adaptability and high yield potential¹⁹.

Plant Disease Detection

Molecular and Spectroscopic Techniques

Indian farming is significantly affected by plant diseases that cause major production and economic losses as well as negative agronomic impact on Indian agriculture²⁰. Plants infected with disease or not, are confirmed by using disease detection molecular techniques. Molecular techniques available for plant disease detection are polymerase chain reaction (PCR), enzyme-linked immunosorbent assay (ELISA), protein sequencing and nucleic acid (DNA) sequencing etc^{21,26} as shown in figure 2.

Plants diseases have been detected by applying unique disease monitoring methods, which are known as the spectroscopic and imaging techniques. Development of such technologies to implement an experimental tool for a large scale disease monitoring in the field, has been involved in current research activities. Symptomatic or asymptomatic plants diseases have been detected by various spectroscopic and imaging techniques. Some methods: fluorescence imaging^{27,29}, multispectral or hyperspectral imaging^{30,32}, infrared spectroscopy^{33,34}, fluorescence spectroscopy^{35,37}, visible/multiband spectroscopy^{38,40}, and nuclear magnetic resonance (NMR) spectroscopy⁴¹ have been used for diseases detection as shown in figure 2.

Powdery mildew (*Erysiphe graminis* sp. *tritici*) in wheat (*Triticum aestivum* L.) was identified by using leaf reflectance measurements⁴².



Fig. 1: Powdery mildew disease of Beans

Similarly, Lorenzen and Jensen⁴³ used spectral reflectance measurements to make an early diagnosis of symptoms on barley leaves infected by cereal powdery mildew. However, the reduction in chlorophyll content is not the common point in disease related stress

Sasaki *et al.* (1998) have distinguished diseased cucumber leaves from healthy leaves at an early infection stage, based upon the spectral reflectance of the leaves in the 500, 600 and 650 nm wavebands. The classification performance was 90%⁴⁴.

In spite of availability of these techniques, there is a demand for a fast, sensitive, and selective method for the rapid detection of plant diseases. In the present review paper, molecular methods and spectroscopic methods of plant disease detection are not discussed in detail as a number of review papers are available in the literature, we are focusing on image processing techniques and describing more in details.

Image Processing and Pattern Recognition techniques

Image processing technology plays important role in the agricultural research in detection of plant diseases. Computer image processing technology was developed in the 1960 of the 20th century, but applied in the 1970s in the area of the production and processing of agricultural products²⁰.

In plant sciences, thousands of images are required for experimental analysis like area of lesions on the leaves, area damaged by insects, scoring quantitative and qualitative traits, etc. All these experiments are analysed manually or with individual software packages. It is very time consuming, tremendous amount of works, problem with analytical accuracy of results and subjectiveness rising from different individuals. So to overcome these issues, plant biologist need efficient software like image processing that can automatically extract and analyze significant content.

Some scientists developed new methods based on image processing and pattern recognition

techniques for fast and accurate grading of plant diseases as shown in figure 2. Many research works have been published regarding the advancements of image processing for feature extraction, segmentation, preprocessing and classification.

A novel approach is proposed⁴⁵ for integrating image analysis technique into diagnostic expert system. A CLASE (Central Lab. of Agricultural Expert System) diagnostic model is used to manage cucumber crop. This expert system diagnoses different disorders such as Leaf miner, Powdery and Downey disease by analyzing four image processing phases: enhancement, segmentation, feature extraction and classification. This proposed methodology reduced error and enhance experimental accuracy.

Recent developments in agricultural science have lead to a demand for a new era of automated methods of plant disease detection. It is necessary that the plant disease detection techniques should be rapid, specific, and sensitive for detection of the disease at early stage²⁰.

An automated system has been developed using algorithm such as bounding box method

chain code technique and moment analysis to recognize and classify fungi disease⁴⁶. Severity of Rust disease on Soybean, also measured by Sobel operator to find out spot edge and plant disease severity by calculating the quotient of disease spot area and leaf area⁴⁷. Kaundal *et al.* (2006) proposed a prediction approach based on support vector machines for developing weather based prediction models of plant diseases⁴⁸. They compared the performance of various techniques like conventional multiple regression, support vector machine and artificial neural network (back propagation neural network, generalized regression neural network). It was observed that Support Vector Machine based regression approach is useful for disease management which could be a better description of the relationship between the environmental conditions and disease level.

Back propagation neural network also proposed for recognition of leaves specificity⁴⁹. It was proved that species of a plant can be specified by back propagation neural network using just shape of leaf image. Thinning algorithm and prewitt edge detection were used to find leaf tokens as input to back propagation algorithm. It was concluded that an expert system can be developed

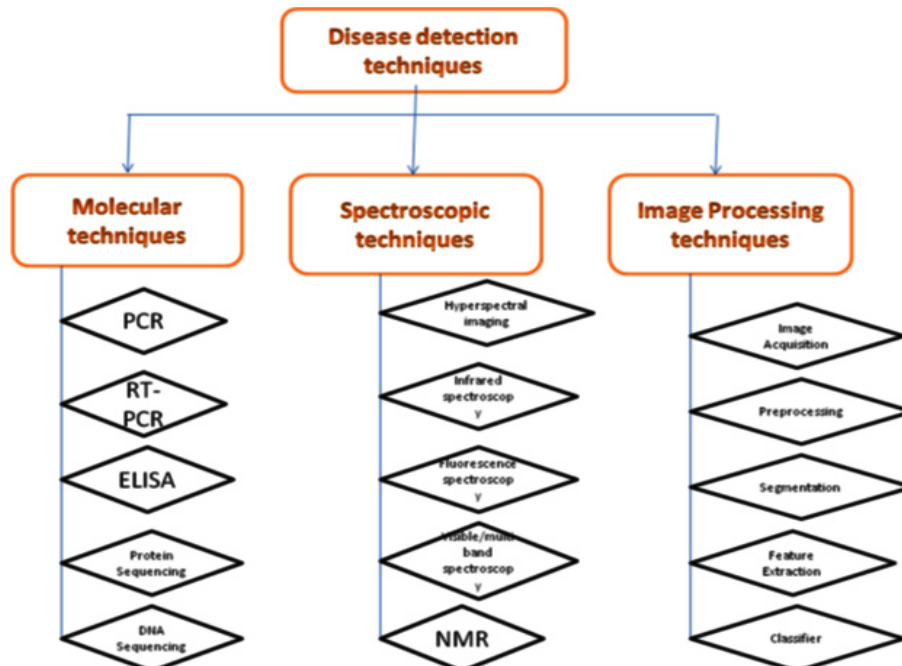


Fig. 2: Flow diagram of Plant Disease Detection techniques

to recognize various plant disease related to leaves like recognition of damaged leaves due to insects and fungal or bacterial diseases of leaves. Neural network approach for segmentation of agricultural lands in remote sensing data was proposed⁵⁰ in which neural network algorithm based on back propagation is used for segmentation of the color images of crop field infected by diseases that changes usual color of plants.

Pixia *et al.*⁵¹ used image processing and recognition technologies for detection of the disease of cucumber downy mildew, powdery mildew and anthracnose. In this study segmentation of leaf spot disease, feature extraction of lesion site, median filtering method of filtering noise were used to detect diseases of images and accuracy rates more than 96% was observed based on the shortest distance methods.

Digital image processing and pattern recognition techniques used for detection powdery mildew disease in the betelvine plants⁵². Data of digital images of the leaves were collected from betelvine plants at various stages using a high resolution digital camera. Analysis of images of the leaves was done using the image processing toolbox in MATLAB. For providing standard patterns of the digital images, preprocessed images were separated using RGB encoding technique the red, green and blue components. These patterns and images of various healthy betelvine leaves at different stages were collected periodically and stored in the system for analysis. Based on these patterns, it was identified whether test leaves were affected by powdery mildew disease or not. Finally using pattern recognition techniques, the powdery mildew disease can be identified at early stage before it spreads to entire crop.

A method was presented to monitor plant disease which caused by spores⁵³. For analysis and processing, the color image was first converted in to gray image (the gray-level correction, histogram generation, feature extraction, image sharpening etc.). Then preprocessing of grey image was done by edge enhancement using the Median filter and canny edge algorithm in order to remove low frequency components. After thresholding binary image obtained was processed by using

morphological features like dilation, erosion, opening etc. It was found that using this method, microscopic plant diseases, optics stripe counting and the chromosome counting etc. can be easily detected. Seiffert *et al.* (2005) also developed new pattern recognition software (HyphArea) for automated quantitative analysis of hyphal growth rates of powdery mildew fungi on plant. By using this software, the development of fungal pathogens can be quantified easily at the level of spore germination or penetration. However, the exact quantification of hyphal growth rates after initial, successful host invasion is much more difficult⁵⁴ but this pattern recognition tool can be helpful for disease detection at early stages.

After studying these methods and techniques, we can draw some conclusions that various molecular and hyper spectral remote sensing techniques are used to identify and detect plant diseases but there are more studies are needed to explore the spectral response characteristics of crops under disease stress of various levels in field and weather conditions. The image processing and pattern recognition techniques could be used to detect disease of crops by feature extraction of infected images and healthy images to distinguish there features on the basis of shape, texture and colour.

On the basis of these techniques, powdery mildew disease can be identified by following features:

- At initial stage, white powdery fungus grows on the upper leaf surface of the lower leaves
- Next stage, fungus spread in the form of mats is called mycelium and appears as white or grayish cottony patches on leaves, buds and stems.
- Final stage, fruiting bodies (cleistothecia) appear as small black or brown specks on the mycelia mats.

CONCLUSION

In this review, there are so many methods used to identify, classify and detect plants disease and using image processing and pattern recognition techniques, preventive action can be taken well in advance such that the entire plantation can be

protected before spreading of the disease. The method of detecting plant disease should be cost effective and non-destructive. Image processing techniques fulfil these characteristics as it not only requires the simple image data of the leaf samples, but also eliminates subjectivity of traditional methods and human induced errors. Early stage detection will helps to farmers to decide the specific quantity for fungicide application which reduces the cost and environmental pollution.

However, a higher number of experiments and improvements in the image-processing subroutine are needed to validate the method. In future we can utilize other principal component classifier such as neural network, back propagation and Support Vector Machine. These analysis methods and sensor specificities can be transferred and generalized for other plant-pathogen systems. Moreover, the technologies may be also used in plant pathology for investigating the effect of pathogenesis on the initial level.

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