

# Automated Leaf Disease Diagnosis with Deep Learning

## Sakshi Abhale<sup>1</sup>, Shrutika Bhavsar<sup>2</sup>, Kadambari Ghadge<sup>3</sup>, Akash Sale<sup>4</sup>

<sup>1</sup> Electronics & Telecommunication, Sinhgad Institute of Technology and Science, Narhe, Pune <sup>2</sup>Electronics & Telecommunication, Sinhgad Institute of Technology and Science, Narhe, Pune <sup>3</sup>Electronics & Telecommunication, Sinhgad Institute of Technology and Science, Narhe, Pune <sup>4</sup>Electronics & Telecommunication, Sinhgad Institute of Technology and Science, Narhe, Pune

Abstract - Agriculture is crucial for India's economy, but one problem is identifying plant diseases, especially in remote areas where contacting experts is difficult. Wrong disease identification can lead to crop loss. Even government agriculture departments sometimes give incorrect advice to farmers. To tackle this, a dataset of apple crop images has been created to help identify healthy and diseased leaves accurately. These images come from different angles and directly from apple fields. The early disease detection in agriculture, particularly focusing on diseases affecting apple crops. Specifically, it mentions diseases like early blight, late blight, black scurf, common scab that can affect apple plants. It discusses using automatic methods like Convolutional Neural Networks (CNN) to detect and classify these diseases based on a dataset. The approach involves automatic feature extraction and classification using colur information.

## **1.INTRODUCTION**

The "Automated Leaf Disease Diagnosis with Deep Learning" project utilizes smart technology, specifically Convolutional Neural Networks (CNN), to accurately distinguish between healthy and diseased apple leaves. By analyzing images from plots, the CNN algorithm detects diseases reliably, allowing farmers to take timely action to protect their plants. This approach not only assists in disease identification but also reduces farming costs by preventing further damage to crops. Project that helps farmers detect diseases in apple crops using smart technology. They use a computer program called a Convolutional Neural Network (CNN) to analyze images of apple leaves and accurately identify any diseases. This helps farmers take quick action to protect their plants, reducing costs and promoting healthier harvests. Ultimately, the project aims to enhance farming practices, promote sustainable agriculture, and contribute to global food security by leveraging technology automated advanced for disease identification in apple crops.

## 2. Literature Review

Paper	Technology	Methodology	Accuracy
No	Used		
[1]	Machine Learning	They used a DL approach based on the Google Net Inception structure and the rainbow concatenation for apple leaf disease detection. They generated a custom dataset of 2,029 images of diseased apple leaves and trained their algorithm to detect five common apple leaf diseases: Alternaria leaf spot, brown spot, mosaic, grey spot and rust [1]	78.80%
[2]	Machine Learning	grey spot, and rust.[1] The goal of Huang et al was to develop a decision support system for spraying machines that could identify Helminthosporium leaf spots in wheat fields using remote sensing data from a UAV. The dataset was divided into four categories based on disease severity: normal, light, medium, and heavy. In the preprocessing step, they extracted 100 × 100 samples from the images.[2]	91.43%
[3]	Deep Learning	Karlekar and Seal proposed a computer vision method to	98.14%



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r		1		1		, 
	identify and categorize				divided into 15	
	leaf diseases present in				classifications, is used	
	soybean crops. The				to help with plant	
	proposed approach				disease recognition	
	separates the leaf				study.[5]	
	section by removing		[6]	Deep	Oppenheim and Shani	96%
	the complex		[0]	Learning	[20] used a CNN	5070
	background from the			Learning		
					based on VGG [21] to	
	entire image.				classify potatoes into	
	Following this, a deep				five classes, four	
	learning Convolutional				diseased ones and a	
	Neural Network				healthy potato class.	
	(CNN) known as Soy				They acquired 400	
	Net, trained on a 16-				images of	
	class dataset called				contaminated potatoes	
	PDDB.[3]				using three simple	
[4] Deep	Mohanty et al trained	99%			digital cameras and	
Learning	-				augmented data by	
	Net for detecting 26				flipping and cropping	
	diseases on 14 crop				the images. In the	
	species using the Plant				e	
	Village dataset with				preprocessing step, they downsized	
					5	
	54, 306 images. In the				images to $224 \times 224$	
	preprocessing step,				and converted to	
	they downsized				grayscale.[6]	
	images to $256 \times 256$ .		[7]	Machine	Khalili et al. [68]	96.79%
	They also			Learning	experimented with six	
	experimented with the				different ML	
	original Plant Village				algorithms, namely	
	dataset, with a grey-				MLP, Gradient Tree	
	scaled version of this				Boosting (GBT), L1	
	dataset, and a version				regularization (LR-	
	in which the leaves				L1), L2 regularization	
	were segmented.[4]				(LR-L2), SVM, and	
[5] Deep	P Vishnu, K Sangeetha	96.63%			RF in order to forecast	
Learning	-	50.0370			charcoal rot disease in	
Loaming	, Sanjutha et.al,				soybean on a custom	
	-				-	
	examined the problem				dataset [69] of 2,000	
	of plant disease				healthy and diseased	
	detection, which is		101		plants.[7]	
	done visually for		[8]	Machine	Author introduced a	98.4%
	identification of plant			Learning	CNN model with eight	
	disease. Compared to				hidden layers that is	
	other forms of				lightweight and	
	photographic photos,				designed for	
	plant disease images				classifying nine types	
	are more likely to				of diseases in tomato	
	feature randomly				crops. The model	
	spread lesions,				outperformed	
	variable symptoms,				traditional ML	
	and complicated				techniques and pre-	
	backdrops, making				trained models with an	
	discriminative in-					
	formation difficult to				accuracy of 98.4% on	
	capture. Plant Village				the publicly available Plant Village dataset,	
1 1	i cadiure. Plant village		1	1	L PLANT VILLAGE dataget	
	Dataset, which contains 20,000 photos				which had 200–1400 images in different	

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		classes.[8]	
[9]	Deep Learning	Wang et al trained and fine-tuned four CNN models, for detecting apple leaf black rot. The authors utilized the Plant Village dataset, which contained two classes of images: healthy apple leaves and apple leaf black rot. The images were further categorized by botanists into four stages: healthy, early stage, middle stage, and end stage.[9]	90.4%
[10]	Deep Learning	Author introduced a multi-stage CNN architecture that was based on AlexNet, aimed at identifying diseases in rice plants. They preprocessed the images by resizing them to 512×512 pixels and then applied the ZCAWhitening technique to remove the correlation between data.[10]	

## **Problem Statement:**

To develop an advanced deep learning system that can accurately detect and diagnose plant leaf diseases by leveraging cutting-edge computer vision techniques and deep neural networks for early disease identification to enhance plant health and agricultural productivity.

## **Experimentation Platform:**

The experiment was done using Google Collab, which is an online platform for coding, and a dataset that had pictures of apple leaves with five types of diseases, plus some pictures of healthy leaves.

Language: Python

Algorithm: CNN

Library: Matplotlib, NumPy, Pandas, Tensor Flow

#### Summary

This project helps farmers by improving crop yields and managing diseases. It identifies problems early on, which helps increase productivity. In the future, it can detect diseases in multiple crops and use data to predict and prevent diseases. Farmers can access this through easy-to-use apps, making it simpler to identify and address plant diseases.

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