

Machine Learning-Based Precision Agriculture

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Abstract- Machine learning-based precision agriculture leverages advanced algorithms to optimize farming practices, improve crop yield, and enhance sustainability. By analyzing diverse datasets such as soil health, weather patterns, crop conditions, and pest presence, machine learning models provide actionable insights for farmers. Techniques like image processing, regression, and classification help in disease detection, irrigation management, and yield prediction. The system delivers real-time recommendations intuitively, empowering farmers to make data-driven decisions, reduce resource consumption, and minimize environmental impact. This approach represents a transformative shift toward more efficient, sustainable, and technology-driven agriculture.

Keywords- Machine learning, Precision agriculture, Crop yield, Sustainability

I. INTRODUCTION

The information technology of the farming sector is currently regarded as a concern to confront the many difficulties that arise. The development of efficient and more profitable agriculture systems and instruments is quickly increasing with the environment. Precision agriculture and intelligent agriculture can lead in this direction. These two phrases concern the integration, with traditional agricultural methods, of sophisticated technologies for the production of good crops [2]. Intelligent agriculture systems can supply farmers with significant environmental data from their fields to enhance competitiveness and profit. Nearly every area of agriculture can benefit from such technical developments, from plantation and irrigation to plant protection and harvesting. When the AI is integrated into the cloud, processes become intelligible and judgments involving human involvement are made fluently [3, 4]. Diverse strategies for resolving contemporary issues in agriculture have been presented, from databases to predictive analysis. As far as precision and efficiency are concerned, systems using AI have proved to be the best results [5]. Farming is a vibrant area, and the circumstances to deliver a meaningful solution cannot be generalized. The AI approaches enable us to collect and respond optimally to the unique challenge of the complicated characteristics of each case. The development of various AI systems

gradually solves extremely complex problems. In agriculture, multiple approaches to agriculture are adapted quickly to AI. The concept of smart systems enables farmers to recognize crops, analyse the soil, provide expert advice, and develop business opportunities. This leads to stochastic AI technologies that enable agricultural production to recognize, collect, and respond to different conditions (depending on the knowledge gained) to increase efficiency [5, 6]. By following new progress in the agricultural sector, farmers can offer solutions via platforms such as chatterbots that benefit from the field. Global agriculture artificial intelligence is predicted to increase significantly [7, 8]. The approach is to detect and control cotton leaf diseases and motivate to increase the application in agriculture-based applications.

II. LITERATURE REVIEW

Title: A Study on Various Techniques for Plant Leaf Disease Detection Using Leaf Image

Author: Sakshi Raina, Abhishek Gupta

Year: 2021

Description: Agriculture plays an important role in the economy of any country and there are a lot of varieties of crops for farmers. The problem or issues occurs when the crops are infected by some disease and the farmers do not know about that disease of plants at the right time. And when the disease is detected, the farmers do not know which disease it

is. Therefore, the examination of automatic leaf disease detection in agriculture is a fundamental subject of research as it could display advantages in the observation of vast fields of yield and thus identify manifestations of disease as they occur on plant leaves. The study of plant disease means the study of different patterns visible with the eyes above the plants' leaves. By looking at the different color and texture features of the same plant, now it can be analyzed which portion of the plant is healthy and which part of the plant has a disease. The process of knowing the disease of the plant occurs in the laboratory. This process takes a lot of time and it is very expensive. For that, the researcher used different types of techniques so that disease would be detected on time and expenses would be reduced. So, this research work attempts to describe the approach suggested by the study articles. Different scholars view the images in terms of Artificial Intelligence and machine learning and demonstrate their achievements and problems that still exist. To draw some assumptions, our study of the various approaches suggested is also given. Image Acquisition, Image Preprocessing, Image segmentation, Feature Extraction and Statistical Analysis, and Classification based on classifiers are the key steps for the identification of diseases. This paper provides, along with the available datasets, a survey of the available approaches to solving the problem discussed.

Title: Identification of Plant-Leaf Diseases Using CNN and Transfer-Learning Approach

Author: Sk Mahmudul Hassan, Arnab Maji

Year: 2021

Description: The timely identification and early prevention of crop diseases are essential for improving production. In this paper, deep convolutional neural network (CNN) models are implemented to identify and diagnose diseases in plants from their leaves, since CNNs have achieved impressive results in the field of machine vision. Standard CNN models require a large number of parameters and higher computation costs. In this paper, we replaced standard convolution with depth=separable convolution, which reduces the parameter number and computation cost. The implemented models were trained with an open

dataset consisting of 14 different plant species, 38 different categorical disease classes, and healthy plant leaves. To evaluate the performance of the models, different parameters such as batch size, dropout, and different numbers of epochs were incorporated. The implemented models achieved disease-classification accuracy rates of 98.42%, 99.11%, 97.02%, and 99.56% using InceptionV3, InceptionResNetV2, MobileNetV2, and EfficientNetB0, respectively, which were greater than that of traditional handcrafted-feature-based approaches. In comparison with other deep-learning models, the implemented model achieved better performance in terms of accuracy and it required less training time. Moreover, the MobileNetV2 architecture is compatible with mobile devices using the optimized parameter. The accuracy results in the identification of diseases showed that the deep CNN model is promising and can greatly impact the efficient identification of diseases, and may have potential in the detection of diseases in real-time agricultural systems.

III. SYSTEM ANALYSIS

3.1 Proposed System

The proposed system utilizes machine learning algorithms to optimize crop management by analyzing data such as soil quality, weather patterns, crop images, and pest information. Image processing and classification techniques are applied for early disease detection and crop health monitoring. Regression models predict soil conditions and crop yield, while also guiding irrigation and fertilization schedules. The system provides real-time, actionable insights through a user-friendly interface, empowering farmers to make data-driven decisions that support sustainable farming practices.

Advantage

- Data Collection
- Increases Land Value
- Increased Product Quality
- Eases Harvesting
- Reduced Costs and Increased Profitability

3.2 Existing system

The problem or issues occurs when the crops are infected by some disease and the farmers do not

know about that disease of plants at the right time. And when the disease is detected, the farmers do not know which disease it is. Every year the production of cotton is reduced due to the attack of the disease. Plant diseases are generally caused by pest insects and pathogens and decrease productivity to a large scale if not controlled within time. The deficiency of adequate decision-making by farmers on yield prediction. There isn't any framework in location to suggest to farmers what plants to grow.

Disadvantage

- Lack of Skilled Workers
- Takes Time
- The primary drawback of this type of technology is cost and Lack of Knowledge

IV.SYSTEM ARCHITECTURE

- Use of WSN in Agriculture
- Irrigation Management System
- Farming System Monitoring
- Pest and Disease Control
- Controlled Use of Fertilizers

V. ALGORITHM

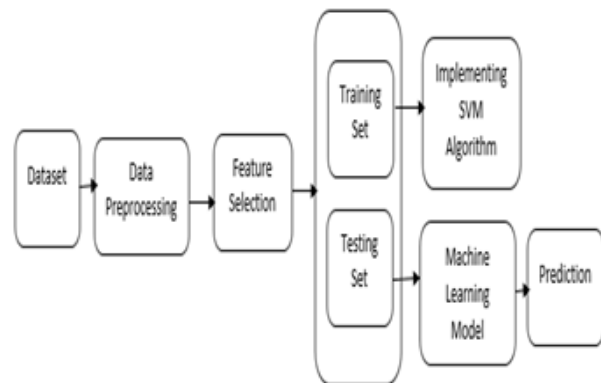
Support Vector Machine Algorithm Most of the tasks machine learning handles right now include things like classifying images, translating languages, handling large amounts of data from sensors, and predicting future values based on current values. You can choose different strategies to fit the problem you're trying to solve.

Supervised vs unsupervised learning Two of the most commonly used strategies in machine learning include supervised learning and unsupervised learning.

Kernel functions

- Linear
- Polynomial
- Gaussian Radial Basis Function (RBF)
- Sigmoid

VI. SYSTEM DESIGN



VII. CONCLUSION

The machine learning-based precision agriculture system presents a significant advancement in optimizing crop management. By analyzing key agricultural data such as soil quality, weather patterns, crop images, and pest information, the system enables early disease detection, improved yield predictions, and more efficient irrigation and fertilization practices. The integration of real-time insights empowers farmers to make informed, data-driven decisions, ultimately enhancing crop productivity while promoting sustainability. This approach fosters a more resilient and eco-friendly agricultural future, aligning with the growing demand for smarter, technology-driven farming solutions.

REFERENCE

1. H. Sharma, A. Haque, and Z. A. Jaffery, "Smart agriculture monitoring using energy harvesting Internet of Things (EHIoT)," *An International Scientific Journal*, vol. 121, pp. 22–26, 2019.
2. M. Suchithra, "Sensor data validation," *International Journal of Pure and Applied Mathematics*, vol. 119, no. 12, pp. 14327–14335, 2018.
3. P. Joshi, "Wireless sensor network and monitoring of crop field," *IOSR Journal of Electronics and Communication Engineering (IOSR-JECE)*, vol. 12, no. 1, pp. 23–28, 2017.
4. S. G. Sangeeta, "Design and implementation of crop yield prediction model in agriculture,"

- International Journal of Scientific & Technology Research, vol. 8, no. 1, 2020.
5. Y. Mekonnen, "Review—machine learning techniques in wireless sensor network based precision agriculture," *Journal of the Electrochemical Society*, vol. 167, no. 3, article 037522, 2020.
6. R. Ghadge, "Prediction of crop yield using machine learning," *International Research Journal of Engineering and Technology*, vol. 5, no. 2, pp. 2237–2239, 2018.
7. G. S. Sujawat, "Application of artificial intelligence in the detection of diseases in plants: a survey," *Turkish Journal of Computer and Mathematics Education*, vol. 12, 2021.
8. P. Kshirsagar and S. Akojwar, "Optimization of BPNN parameters using PSO for EEG signals," in *Proceedings of the International Conference on Communication and Signal Processing*, pp. 385–394, India, 2016.
9. S. Oza, A. Ambre, S. Kanole, et al., "IoT: the future for quality of services," in *ICE 2020* pp. 291–301, Springer, Singapore.
10. P. K. Kollu, K. Kumar, P. R. Kshirsagar, et al., "Development of advanced artificial intelligence and IoT automation in the crisis of COVID-19 Detection," *Journal of Healthcare Engineering*, vol. 2022, Article ID 1987917, 2022.
11. P. R. Kshirsagar, P. P. Chippalkatti, and S. M. Karve, "Performance optimization of neural network using GA incorporated PSO," *Journal of Advanced Research in Dynamical and Control Systems*, vol. 10, no. 4, pp. 156–169, 2018.
12. N. Ahmed, D. De, and I. Hussain, "Internet of Things (IoT) for smart precision agriculture and farming in rural areas," *IEEE Internet Things Journal*, vol. 5, no. 6, pp. 4890–4899, 2018.
13. S. Sundaramurthy, C. Saravanabhavan, and P. Kshirsagar, "Prediction and classification of rheumatoid arthritis using ensemble machine learning approaches," in *2020 International Conference on Decision Aid Sciences and Application (DASA)*, pp. 17–21, India, 2020.
14. M. Padmaja, S. Shitharth, K. Prasuna, A. Chaturvedi, P. R. Kshirsagar, and A. Vani, "Grow of artificial intelligence to challenge security in IoT application," *Wireless Personal Communications*, 2021.
15. Sarangdhar and V. R. Pawar, "Machine learning regression technique for cotton leaf disease detection and controlling using IoT," in *International Conference on Electronics, Communication and Aerospace Technology*, pp. 449–454, India, 2017.