Meta-Analytical Approach of ML, IoT and Nanotechnology for Plant Disease Detection towards a Sustainable Agriculture

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Abstract— This review article examines various technologies, such as Machine Learning, the Internet of Things (IoT), and Nanotechnology, that have the potential to address plant disease detection issues and provide sustainable long-term solutions. In addition, how these strategies can be integrated into precision agricultural practices to enhance crophealth and productivity has also been discussed and analyzed. This meta-analysis aims to contribute to the advancement of agriculture and facilitate informed decision-making by stakeholders in the industry. A thorough understanding of the present landscape of plant disease detection and the challenges that persist can lead to innovative solutions that guarantee the long-term viability of agriculture. This presentation will provide an in-depth overview of current research, highlighting the latest technological advancements and strategic approaches that have the potential to revolutionize the identification and control of plant diseases in agricultural systems.

Index Terms—plant disease detection, sustainable agriculture, machine learning (ML), Internet of Things (IoT), Nanotechnology

INTRODUCTION

The increasing global population and corresponding rise in food consumption present unprecedented challenges for the agriculture industry. In response, agricultural practices have advanced significantly by choosing high-yielding cropvarieties, precision agriculture, and modern irrigation systems. However, as the pursuit of greater agricultural output continues, a challenging adversary has emerged: plant diseases. These microscopic opponents, comprisingbacteria, fungi, viruses, and pests, have long erodedagricultural yields and quality. This has led to significant economic losses and food scarcity and has consequently spurred substantial research into plant disease detection and control. The objective of this review paper is to evaluate the current status of plant disease detection, identify the obstacles to progress, and provide long-term solutions. Plant diseases can significantly reduce crop yield and quality, leading to economic losses and trade restrictions in agriculture. Pesticide use harms the environment and poses health risks. We must take this issue seriously and findeffective solutions to protect our food supply and our planet's health. Climate change has led to more aggressive plant diseases, making traditional management strategies ineffective. It's crucial to adopt proactive measures for

efficient disease management. There are traditional and contemporary methods to identify plant diseases, which have

been refined over time and are effective. Farmers have relied on visual inspection, symptom monitoring, and symptom-based guidance for decades. These methods have limitations, including subjectivity, reliance on human expertise, and inability to detect latent infections or distinguish between similar symptoms. Despite advances in plant disease identification, the agricultural industry still faces challenges in ensuring sustainability. Rapid and precise detection technologies are crucial to limit the spread of disease, but current approaches often fall short. Affordability is also a significant barrier for many farmers.

Plant diseases have far-reaching impacts, beyond just agriculture. The economic losses, environmental impact of pest control technologies, and trade restrictions highlight the significance of having long-term and efficient solutions for the detection and management of plant diseases. The use of chemical treatments in a reactive manner poses environmental risks and raises concerns about the sustainability of current farming practices.

I. LITERAURE SURVEY

A recent study conducted by H.F. Pardede, E. Suryawati, D. Krisnandi, R.S. Yuwana, and V. Zilvan [1] has demonstrated that Deep Convolutional Neural Networks (CNN) outperform traditional techniques such as Support Vector Machine (SVM)

in detecting plant diseases. Machine-learning algorithms improve disease detectionaccuracy by automatically extracting essential features from raw data. However, the field faces two significant challenges: resilience to environmental variations and dealing with a vast variety of data and diseases with limited datasets. Addressing these challenges is crucial for future

advancements in the sector and realizing the potential of machine learning for sustainable agriculture.

The study by U. Shruthi, V. Nagaveni, and B. K. Raghavendra [2] brings to the fore a pressing need to counteract the decline in crop yields due to bacterial, viral, and fungal diseases. The study recommends using data- driven machine learning algorithms for disease diagnosis in agriculture. It highlights the importance of higher agricultural output to support India's growing population. The study shows that convolutional neural networks (CNNs) are more accurate in detecting various plant diseases. However, practical challenges, data limitations, and long-term sustainability must be considered for successfulimplementation.

The review by Chenghai Yang [3] underscores the critical importance of leveraging aerial and satellite imagery forearly detection and control of diseases in crops throughout the growing season, particularly in the context of precision agriculture. The study examines the use of remote sensing technology for diagnosing and managing agricultural diseases, as well as preventive measures. The study shows how variable rate technology can apply fungicides to infected areas, enhancing disease control. While remote sensing has potential, further research and standardization are needed to optimize effectiveness.

The issue of plant diseases and their impact on sustainable agricultural output and global nutritional sustainability was the focus of a research conducted by Younas et al. [4] The authors highlighted that these diseases have significant effects, resulting in economic losses and lower yields for various cash crops. Effective disease management techniques are crucial for sustainable agricultural productionand food security. A thorough study and analysis of these measures are necessary to optimize land usage, minimize input costs, and address environmental concerns.

The study carried out by Tahir Farooq, Muhammad Adeel, Zifu He, Muhammad Umar, Noman Shakoor, Washingtonda Silva, Wade Elmer, Jason C. White, and Yukui Rui [5] presents a novel approach to addressing the challenges posed by phytoviruses in the field of agriculture, namely the discipline of "Nanophytovirology". Synthetic nanomaterials have potential in treating plant diseases but their efficacy against phytoviruses is unclear. These organisms have distinct physiochemical features that enable unique interactions with viruses, vectors, and host

plants. As a result, it presents an opportunity for novel approaches in managing viral diseases. Further multidisciplinary research is imperative to fully unlock the potential of nanotechnologyin effectively diagnosing, treating, and preventing viral diseases.

Another review article by S. R. Prathibha, A. Hongal and M.

P. Jyothi [6] stresses on the critical importance of the Internet of Things (IoT) in advancing the smart agriculture paradigm, particularly in the fields of automation and environmental monitoring. IoT sensors like CC3200 chip can increase agricultural yields by providing real-time data on crucial environmental factors. The report proposes new research avenues to address scalability issues and broaden the scope of monitored factors for greater efficiency and sustainability in farming.

The review article presented by I. M. Marcu, G. Suciu, C.

M. Balaceanu and A. Banaru [7] offered an overview of how technology, particularly the Internet of Things (IoT), has transformed agriculture, revolutionizing the old practices. The article discusses the evolution of agriculture, from manual labor to IoT technologies for precisionagriculture. The study proposes an IoT-based system that monitors crucial factors affecting crop health, improving accuracy and productivity. The technology has the potential to transform agriculture but requires further research to address challenges such as scalability and data integration. The study serves as a stepping stone for future research, highlighting the need to make agriculture more robust and efficient to address current environmental challenges.

Another review conducted by S. Suhag, N. Singh, S. Jadaun,

P. Johri, A. Shukla and N. Parashar [8] shed light on the challenges that farmers encounter in meeting the food requirements of a growing population. The report highlights the significance of innovative technologies for better agricultural productivity and food security. The proposed framework using IoT technology integrates sensors to collect real-time data on plant health, soil fertility, water quality, and temperature. The development of a harvesting robotic arm with image recognition capabilities is a significant step forward in agricultural automation. However, further research is necessary to optimize technology for practical use and address existing gaps indata integration, scalability, and interaction with the agriculture community.

The review article by Jian Yang, Amit Sharma, Rajeev Kumar [9] emphasizes the critical importance of agriculture in a country's progress, with a particular focus on India, where the vast majority of the people relies on agriculture for their livelihood. The article emphasizes the need for modernizing agriculture through smart farming techniques, which use sensors

and the IoT. The proposed framework involves smart sensors for data collection and cloud-based analysis. The benefits of this system include precision control, automatic monitoring, and datarich insights, potentially promoting sustainable agriculture practices.

This comprehensive review study by Bryony E.A. Dignam, Maureen O'Callaghan, Leo M. Condron, Jos M. Raaijmakers, George A. Kowalchuk, Steven A. Wakelin

[10] underscores the critical role of grasslands in preserving biodiversity and providing essential ecosystem services, particularly in the context of pastoral agriculture. Soil-borne microbial diseases pose a challenge to pasture productivity, complicated by the interactions within the soil microbiome. In pastures, controlling plant infections is difficult due to limited options and diverse pathogen complexes. Disease suppression can be achieved through soil microbiota and understanding its traits. A systems-based approach that considers microbial features and ecosystem complexity is crucial to identify disease suppressive mechanisms and improve sustainable control of plant pathogens in grazing systems.

Another review study by Paul Vincelli [11] highlights the potential of genetic engineering (GE) to enhance disease resistance in agricultural plants, promote sustainable agriculture practices, and reduce pesticide use. The study highlights the importance of evaluating individual cases based on risks, benefits, and societal factors when deploying GE traits. It emphasizes a long-term approach and careful implementation for sustainability. While exploring risks and

challenges associated with GE, the study recognizes its potential for precise and targeted genetic modifications as a valuable supplement to traditional breeding methods. Making educated decisions about its use in agriculture is necessary, considering the constantly evolving landscape of genetic engineering.

A research review conducted by Fuentes A, Yoon S, Kim SC, Park DS [12] focused on a strategy that uses deep-learning to detect plant diseases and pests in tomato plants at an early stage. The researchers tested deep learning meta- architectures and developed a new technique for annotation and data augmentation to improve accuracy and reduce falsepositives. A recent study using a Tomato Illnesses and Pests Dataset confirmed deep learning's effectiveness in recognizing nine illnesses and pests in complex plant surroundings. Such technologies have the potential to reduce agricultural losses and promote sustainable farming practices, but further research is needed to improve scalability and real-time applicability.

Another review analysis by Arti Singh Baskar, Ganapathy Subramanian, Asheesh Kumar Singh and Soumik Sarkar

[13] highlights the significance of machine learning (ML) in

tackling the "big data" problem of high-throughput phenotyping (HTP) in plant sciences. This review provides adiverse range of machine learning approaches for plant science, including identification, classification, quantification, and prediction of biotic and abiotic stress. It also highlights the need for further research on ML's applicability to high-resolution imaging and plant stress analytics to enhance its potential in agriculture.

Rising global population demands higher crop productionyield while mitigating the impact of pests and diseases. Tiago Domingues and João C. Ferreira [14] conducted a review of the use of Machine Learning (ML) techniques in agriculture for pest and disease detection, classification, and prediction. Machine learning can enhance crop productivity and reduce the impact of pests and diseases on agriculture. Further research is needed to explore its integration in various

agricultural contexts, developing tailored models for specific challenges associated with different crops and regions to promote sustainable and efficient practices globally. convolutional neural networks (CNNs), to identify plant diseases.

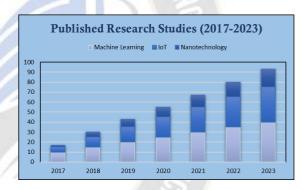


Fig 1. Trend in Research Studies Related to Plant Disease Diagnosis Using Nanotechnology, IoT, and Machine Learning (2017-2023)

The main focus of this meta-analysis is on a classification parameter based on type of diseases in plants. The study includes fungal, bacterial, viral, parasitic and diseases based on nutritional disorders. This study aims to address the current challenges in plant disease detection while anticipating a future of sustainable, resilient, and technologically advanced agriculture.

A. Proposed Framework

- Out of three mainly focused techniques Machine Learning is something which can be used throughout the detection process for feature and symptom analysis.
- On the other hand, Internet of Things and nanotechnology can be used for data acquisition using sensors and chips to identify whether the part(s) of plants are infected or not.

• Steps of implementation of machine learning with IoT or Nanotechnology are shown in *Fig.* 2

II. SYNTHESIS OF FINDINGS

In this review paper, we have researched and collected information from more than 50 papers published during the years 2017-2023 from the peer-reviewed journals of different databases such as Scopus and Web of Science analogous to the keywords such as plant disease identification, machine learning, IoT, nanotechnology and sustainable agriculture. *Fig. 1* shows the trend in research studies related to plant disease diagnosis using nanotechnology, IoT and machine learning.

After conducting a comprehensive study, we can confidently state that this review aims to expound on the thoughts and discoveries of those who have researched the complex issue of plant diseases and their diagnosis by delving into the vast body of earlier research. The study specifically focuses on the use of Machine Learning (ML) technology and implementation of ML algorithms, with a particular emphasis on deep learning techniques, such as

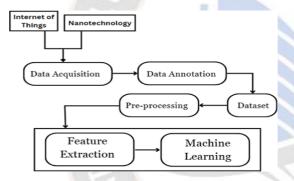


Fig 2. Implementation Steps of Machine Learning Models in Plant Disease Detection.

• Examples along with the types of diseases that can be detected through integration of machine learning with IoT or nanotechnology are shown in *Table*. *1* and *Fig*. 2.

Table I Type of Diseases with Detection Techniques

S. No.	Types of Diseases	Detection Techniques (Integration with Machine Learning)	Examples
2	Bacterial	IoT	Citrus Canker
3	Viral	IoT and Nanotechnology	Tomato Yellow Leaf Curl Virus (TYLCV)
4	Parasitic	IoT and Nanotechnology	Iron deficiency chlorosis, nitrogen deficiency, phosphorus deficiency
5	Nutritional disorder	IoT	Root Knot Nematodes, Cyst Nematodes, Aphids

B. Meta-Analysis of Proposed Approaches

While ML has the ability to automate detection processes and analysis of features, there are recurring obstacles that need

attention. The study discusses IoT and its use in smart farming, nanotechnology for disease detection, and remote sensing technologies. It provides an overview of technology developments, challenges, and future opportunities in plant disease detection. Machine Learning (ML) techniques are revolutionizing the detection of plant diseases. These algorithms can identify diseases quickly and effectively by recognizing complex patterns and trends from raw data. Machine learning excels at adapting to different datasets and learning from large amounts of input data. They can revolutionize agriculture by outperforming traditional approaches, particularly in plant disease identification and real-time response to emergent dangers. ML can revolutionize sustainable agriculture, but it faces challenges. Adapting to environmental changes and managing diverse data is crucial for successful implementation.

Farmers can monitor temperature, humidity, and soil moisture in real-time with IoT technology for better plant health. This helps identify early signs of disease, preventing further damage, and making informed decisions about the use of pesticides, thus enhancing productivity. The use of machine learning and IoT can help gather and analyse contextual data in agriculture. This transformational technology can increase crop yields, reduce costs, and eliminate the use of hazardous pesticides. However, to fully realize the benefits of smart agriculture, dedicated research must address challenges such as scale, data integration, and restricted access to technology.

Nanotechnology has vast potential in disease diagnosis and management, including Phyto virus control. It has the powerto revolutionize agriculture, but practical application needs further investigation. Multidisciplinary research is required to bridge the gap between theory and real-world use. Innovative solutions are essential to address scalability and implementation challenges. By integrating Machine Learning, the Internet of Things, and Nanotechnology, we can now identify plant diseases more accurately and efficiently. As shown in *Fig. 3.*, this convergence marks a new era in agriculture, promising to revolutionize disease detection and contribute to sustainable food production.

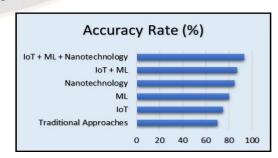


Fig 3. Graph illustrating the improvement in disease detection accuracy with integrated technologies.

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C. Gap Analysis

There are several issues to consider when examining the gap between current plant disease detection technologies and their potential for sustainable agriculture. Although MachineLearning (ML), the Internet of Things (IoT), and Nanotechnology have the potential to improve disease detection efficiency, there is still a significant obstacle of high implementation costs. This cost constraint is a major barrier, particularly for farmers who are resource- constrained, preventing widespread access to these cutting- edge technologies and hindering their full integration into agricultural operations. In addition, despite the progress made in these areas, there is still a need to translate scientific innovation into practical solutions that align with agriculture's broader sustainability objectives.

In spite of the promising future of agricultural technology such as robots and artificial intelligence, it is still uncertain whether they can be made accessible and affordable forfarmers. To address this issue, it is necessary to not only develop new technology but also ensure its scalability and applicability to different farming scenarios. Moreover, academics, politicians, and agricultural stakeholders should work together to overcome the systemic barriers that hinder the implementation of breakthrough agricultural technology.

To sum up, we need to use technology to promote sustainable agriculture that is accessible, affordable, and inclusive. This requires comprehensive approaches that address socio-economic, regulatory, and institutional issues. Our goal is to create a future where sustainable agriculture is a practical reality for all farmers.

CONCLUSION AND FUTURE SCOPE

The combination of Internet of Things and MachineLearning has the potential to enhance disease detection accuracy in agriculture. By utilizing the real-time monitoringcapabilities of IoT with the pattern recognition algorithms of ML, the combination can increase the accuracy of illness detection by a few percentage points to 10-20%, according to studies. Moreover, when combined with Nanotechnology, the usage of IoT, Nanotech, and ML can improve accuracy and efficiency in illness detection by 10-30% or more. It is difficult to widely adopt advanced technologies in agriculture due to their high deployment costs. This makes itharder for farmers with limited resources to access them, which in turn threatens sustainable agricultural initiatives. However, despite these challenges, the predicted advancements in robotics and artificial intelligence offer hope for future solutions that are both accessible and cost- effective.

In order to make the most of IoT, Nanotech, and ML in agriculture, more research and innovation are needed to overcome the barriers to their wider use. Efforts should focuson developing affordable and scalable solutions that can be accessible to all farmers, regardless of resource constraints.

Additionally, multidisciplinary collaboration is necessary to fully realize the potential of these technologies and overcomethe socioeconomic, regulatory, and institutional challenges. As technology continues to advance, there will be anopportunity to explore the integration of robots and AI intoillness detection systems. Future research can focus on thepossible uses of these developing technologies, which couldpave the way for more efficient and sustainable farmingtechniques.

In conclusion, the combination of IoT, Nanotech, and ML has the potential to improve disease detection in agriculture. However, it is critical to overcome deployment costs and ensure accessibility for all farmers. By embracing technological breakthroughs and encouraging cross- disciplinary collaboration, we can enable farmers to tackle illnesses and promote sustainable agriculture for future generations.

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