

Review Of Solar Powered Multipurpose Agriculture Robot Using ML

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doi.org/10.64643/IJIRTV12I10-204544-459

Abstract—Agriculture is a vital sector in developing countries, yet it faces major challenges such as labor shortages, inefficient resource utilization, and environmental concerns. This paper presents a review of solar-powered multipurpose agricultural robots integrated with Machine Learning (ML) techniques. These systems automate farming operations such as ploughing, seed sowing, irrigation, pesticide spraying, and crop monitoring. The integration of ML enables intelligent decision-making, improving efficiency and productivity. The use of solar energy ensures sustainability and reduces operational cost. This paper discusses system architecture, working principles, advantages, limitations, and future scope of smart agricultural robots.

Index Terms—Agricultural Robot, Machine Learning, Internet of Things (IoT), Solar Power, Precision Agriculture, Smart Farming, Automation

I. INTRODUCTION

Agriculture plays a crucial role in the economy, especially in countries like India where a large population depends on farming. Traditional agricultural practices involve significant manual labor and lack precision, resulting in low productivity and higher costs.[5]

With the advancement of technologies such as:

- Machine Learning
- Internet of Things (IoT)
- Robotics
- Renewable Energy

modern agriculture is evolving into smart farming systems.

Agricultural robots (agribots) are designed to:

- Reduce human intervention
- Increase efficiency

- Perform repetitive tasks automatically
- Optimize resource usage

The integration of solar power makes these systems eco-friendly and cost-effective.

II. LITERATURE REVIEW

Several research works have been carried out in the field of agricultural robotics:

Agricultural robotics has gained significant attention in recent years due to the increasing demand for automation and precision in farming practices. Various researchers have proposed solar-powered multipurpose agricultural robots to reduce human effort and improve productivity.

One of the early implementations focuses on solar-powered autonomous agricultural robots capable of performing basic farming operations such as seed sowing, pesticide spraying, and grass cutting. These systems utilize photovoltaic panels to generate energy, which is stored in batteries and used to power motors and controllers. The robot is typically controlled using microcontrollers like Arduino and communicates via Bluetooth or Android applications. This approach significantly reduces labor dependency and enhances operational efficiency. However, these systems largely rely on manual commands and lack intelligent decision-making capabilities.[1]

Another important development is the integration of Internet of Things (IoT) in agricultural robots, enabling real-time monitoring and remote operation. These systems employ sensors such as soil moisture sensors, temperature sensors, and PIR sensors to collect environmental data. The collected data is transmitted to cloud platforms like Firebase, allowing farmers to monitor field conditions through mobile

applications. Additionally, features like intrusion detection using cameras and fire detection systems improve field security and safety. Despite these advancements, the decision-making process in such systems is still limited and mostly rule-based rather than learning-based.[3]

A comprehensive survey on solar-powered agricultural robots highlights the use of multipurpose robots capable of performing tasks such as cultivating, digging, seeding, insecticide spraying, and weed removal. These robots are designed to perform multiple operations simultaneously, improving efficiency and reducing time consumption. The use of DC motors, embedded controllers, and programmable mechanisms allows precise control over agricultural operations. The survey also identifies key challenges such as dependency on solar energy, communication limitations, and navigation issues in uneven terrains.[2]

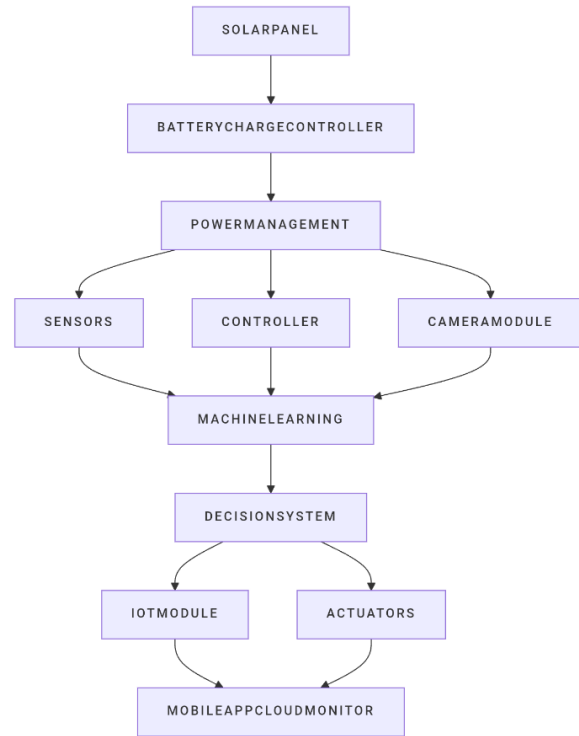
In addition, several systems incorporate image processing techniques for monitoring and security purposes. For instance, cameras integrated with Raspberry Pi are used to capture images of the field, which are then processed to detect intrusions or abnormal conditions. While this approach introduces a level of automation and intelligence, it is primarily based on basic image processing techniques rather than advanced Machine Learning models.[3]

Recent trends in agricultural robotics emphasize the need for Machine Learning and Artificial Intelligence integration. ML algorithms can analyze large volumes of sensor and image data to make accurate predictions regarding crop health, irrigation needs, and pest infestations. Compared to traditional automation systems, ML-based systems offer adaptive learning capabilities and improved decision-making. However, current implementations are still in early stages, and there is a significant gap in developing fully autonomous, intelligent agribots.

Research Gap

- Lack of advanced Machine Learning-based decision systems
- Limited autonomous operation in existing robots
- Heavy dependence on manual or remote control
- Challenges in navigation and obstacle detection
- Insufficient integration of real-time predictive analytics

III. SYSTEM ARCHITECTURE



Components

- 1). Solar Panel
 - Converts sunlight into electrical energy
 - Provides power to the system
- 2). Battery & Charge Controller
 - Stores energy for continuous operation
- 3). Controller Unit
 - Arduino / Raspberry Pi
 - Controls all operations
- 4). Sensors
 - Soil moisture sensor
 - Temperature sensor
 - Humidity sensor
 - PIR sensor
- 5). Camera Module
 - Captures images for ML processing
- 6). Machine Learning Module
 - Detects crop diseases
 - Identifies weeds
 - Supports decision-making

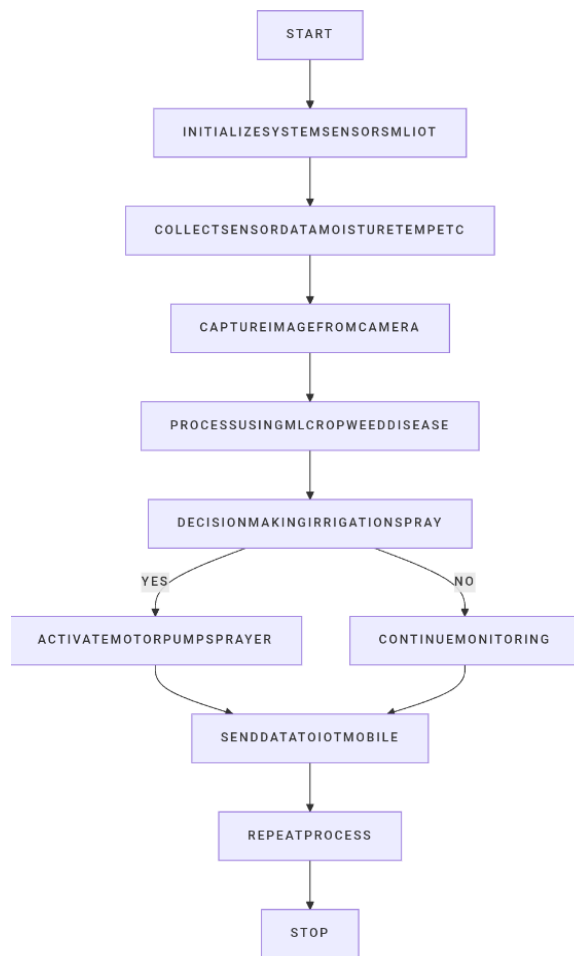
7). IoT Module

- Enables remote monitoring via mobile app

8). Actuators

- Motors (movement)
- Pump (irrigation)
- Sprayer (pesticides)
- Seeder mechanism

IV. WORKING PRINCIPLE



1. System initializes all modules
2. Sensors collect environmental data
3. Camera captures crop images
4. ML model processes the data
5. Decision system determines required action
6. Actuators perform tasks (watering, spraying, etc.)
7. Data is sent to IoT platform
8. Process repeats continuously

V. ROLE OF MACHINE LEARNING

Machine Learning (ML) acts as the cognitive layer of modern agricultural robotics, transforming raw sensor data into actionable intelligence. By shifting from static, rule-based programming to adaptive learning, these systems can respond to the dynamic conditions of a field environment.

1. Crop Disease Detection

Advanced computer vision models allow for real-time health assessments:

- **Image Classification:**
Using Convolutional Neural Networks (CNNs), the robot identifies patterns in leaf color and texture to diagnose specific pathogens.
- **Early Intervention:**
Detecting diseases at nascent stages prevents widespread crop failure and reduces the overall need for chemical treatments [4].

2. Smart Irrigation

ML algorithms move beyond simple moisture thresholds to predictive modeling:

- **Predictive Water Requirements:**
The system analyzes historical soil data and ambient humidity to forecast exactly when a plant will need water.
- **Resource Conservation:**
By providing water only, when necessary, the system prevents water wastage and avoids the nutrient leaching caused by over-irrigation.[2]

3. Weed Detection

Precision application is made possible through automated visual differentiation:

- **Species Differentiation:**
ML models are trained to distinguish the morphological features of primary crops from invasive weeds.
- **Targeted Spraying:**
This enables the robot to apply herbicides exclusively to the weeds, significantly lowering chemical usage and protecting the main crop.[3]

4. Autonomous Navigation

For a robot to be truly multipurpose, it must navigate complex terrains without human oversight:

- Path Planning:

Algorithms determine the most efficient route through a field to minimize energy consumption and soil compaction.

- Obstacle Detection:

Real-time processing of ultrasonic or visual data allows the robot to identify and avoid hazards such as rocks, trees, or workers.[6]

The integration of these ML modules directly addresses the limitations found in basic automation systems, moving agriculture toward a fully autonomous, precision-based model

Additionally, the integration of IoT allows real-time monitoring and remote control, making the system highly efficient and user-friendly for farmers.

However, certain limitations such as high initial cost, dependency on sunlight, and connectivity issues in rural areas still exist. Despite these challenges, continuous advancements in artificial intelligence, energy storage technologies, and wireless communication are expected to overcome these barriers.

In conclusion, solar-powered ML-based agricultural robots have the potential to revolutionize the agricultural sector by transforming it into a smart, efficient, and sustainable system. Future research should focus on enhancing autonomy, improving navigation accuracy, and developing cost-effective solutions to make this technology accessible to small and medium-scale farmers.

VI. APPLICATIONS

1. Precision farming
2. Automated irrigation systems
3. Crop monitoring
4. Greenhouse management
5. Smart village agriculture

VII. CONCLUSION

The development of a solar-powered multipurpose agricultural robot integrated with Machine Learning represents a significant advancement toward modern precision agriculture. This system effectively combines renewable energy, automation, and intelligent decision-making to address key challenges faced in traditional farming such as labor shortages, inefficient resource utilization, and environmental concerns.

The incorporation of Machine Learning enables the robot to perform complex tasks such as crop health monitoring, disease detection, and smart irrigation with minimal human intervention. Unlike conventional automated systems, ML-based agribots can adapt to varying environmental conditions and make data-driven decisions, thereby improving overall productivity and crop yield.

The use of solar energy further enhances the sustainability of the system by reducing dependence on fossil fuels and lowering operational costs.

APPENDIX

A. Hardware Components Specification

Microcontroller- Arduino Uno / Raspberry Pi
Solar Panel -12V, 10W–20W photovoltaic panel
Battery -12V rechargeable battery
Motor Driver -L293D / L298N
DC Motors -12V geared motors
Sensors-SoilMoisture, Temperature, Humidity, PIR
Camera -USB Camera / Pi Camera
Communication -Wi-Fi / Bluetooth (HC-05).

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