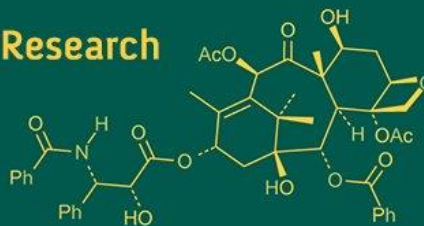


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## Artificial intelligence in agriculture: A pathway toward sustainable and smart farming

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### Abstract

Artificial Intelligence (AI) is transforming agriculture by introducing intelligent systems that improve productivity, resource management, and sustainability. This journal presents an overview of AI applications in agriculture, focusing on irrigation, crop monitoring, disease detection, fertilization, and automated machinery. Drawing on interdisciplinary studies and real-world developments, the paper highlights the impact of AI in solving challenges such as labor shortages, inefficient input use, and climate variability. Additionally, it discusses the barriers to adoption and the potential of AI to support sustainable development goals. This work contributes a structured, accessible overview suitable for students and general readers with a foundational interest in smart agriculture.

**Keywords:** Artificial intelligence, precision agriculture, smart farming, sustainability, crop monitoring, robotics, drones

### 1. Introduction

Agriculture is a key sector that supports global food security, employment, and economic stability. However, traditional farming practices face critical limitations due to increasing population demands, climate change, land degradation, and a shrinking labor force. To address these challenges, technological solutions are being adopted worldwide. Among them, Artificial Intelligence (AI) stands out as a transformative force in modern agriculture.

AI refers to the simulation of human intelligence in machines capable of learning, reasoning, and making decisions. In agriculture, AI allows for the development of smart systems that monitor field conditions, analyze data, and perform tasks automatically or with minimal human intervention. From detecting plant diseases to optimizing irrigation, AI supports more precise, productive, and environmentally friendly farming.

This journal presents a structured review of AI applications in agriculture, based on recent studies and technologies. It aims to explain the role of AI in modern farming using simplified academic language to make the topic accessible to a wider audience.

### 2. Applications of Artificial Intelligence in Agriculture

#### 2.1 Precision Irrigation and Water Management

Efficient water use is essential for sustainable agriculture, especially in regions prone to drought. Traditional irrigation methods often waste water due to overuse or incorrect timing. AI-based irrigation systems use real-time sensor data, weather forecasts, and soil moisture levels to schedule water delivery.

According to Talaviya *et al.* (2020) <sup>[1]</sup>, sensor-based irrigation combined with AI can significantly reduce water consumption while improving crop yield. Systems using the dielectric method and neutron moderation can accurately measure soil water content. AI then analyzes this data to determine the optimal irrigation schedule.

#### 2.2 Weed Detection and Control

Weed control is another major challenge in agriculture. Conventional methods involve manual labor or widespread use of herbicides, both of which have limitations. AI enables the development of robotic weeding systems that distinguish between weeds and crops using computer vision.

A study by Kumar *et al.* (2020) <sup>[3]</sup> introduced a mechatronic prototype using a fuzzy logic algorithm that adjusts weeding blade movement based on plant location and soil conditions. This system reduces crop damage and minimizes herbicide use, promoting eco-friendly farming.

### 2.3 Pest and Disease Detection

Plant diseases and pests are major causes of yield loss worldwide. Early detection is crucial to prevent outbreaks and reduce pesticide use. AI technologies can detect diseases by analyzing leaf images, color changes, and growth patterns using machine learning models.

For instance, Singh *et al.* (2020) <sup>[9]</sup> highlighted how hyperspectral imaging, thermal imaging, and convolutional neural networks (CNNs) are used to classify and detect plant diseases. These systems offer high accuracy in disease identification and can operate both in real-time and offline.

### 2.4 Drone-based Monitoring and Spraying

Drones equipped with AI systems are widely used for crop monitoring, mapping, and targeted spraying. They offer fast, aerial views of large fields and collect data on plant health, water stress, and pest infestation. The integration of AI allows drones to interpret this data and carry out precise actions such as spraying only infected areas.

Talaviya *et al.* (2020) <sup>[1]</sup> describe three types of drone sprayers: hydraulic, gaseous, and centrifugal, all of which benefit from AI for effective delivery and reduced chemical waste.

### 2.5 Fertilizer and Nitrogen Optimization

Excessive use of nitrogen-based fertilizers can lead to soil damage and environmental pollution. AI supports smart nutrient management by analyzing crop requirements, soil nutrient levels, and growth stages.

The concept of Agri-BIGDATA, proposed by Yang *et al.* (2020) <sup>[4]</sup>, integrates sensor data, genetic information, and remote sensing for optimal nitrogen input planning. This reduces waste and promotes sustainable nutrient use across different crop systems.

## 3. AI in Agricultural Equipment and Human Safety

Apart from improving crop production, AI also contributes to ergonomics and safety in agricultural machinery operations. Manual tasks like pedaling, braking, and clutch use on tractors can lead to musculoskeletal disorders. Hota *et al.* (2020) <sup>[6]</sup> developed a foot transducer system that tracks lower limb force and recommends ergonomic improvements.

AI-based automation in machinery also reduces the need for repetitive manual labor. Autonomous tractors and harvesters, guided by GPS and AI, improve precision and reduce human error.

## 4. Computer Vision and Livestock Monitoring

AI also supports precision livestock farming. Using computer vision and deep learning, AI systems can monitor animal behavior, health, weight, and stress levels. Okinda *et al.* (2020) <sup>[10]</sup> reviewed poultry monitoring systems that use image processing to detect lameness, illness, and behavioral patterns.

By automating animal health checks, farmers can reduce losses and improve animal welfare without increasing labor demands.

## 5. Benefits and Impact of AI in Agriculture

The integration of AI in agriculture brings numerous benefits to farmers, consumers, and the environment. These benefits are not just limited to productivity gains but also include economic, environmental, and social improvements.

### 5.1 Increased Crop Yield and Productivity

AI-driven tools allow for precise and timely actions in farming. By using data collected from soil sensors, drones, and weather forecasts, AI can determine the best time to plant, irrigate, fertilize, and harvest. This optimization leads to higher crop yields with lower input costs.

According to the review by Oliveira and Silva (2023) <sup>[8]</sup>, over 20 different AI techniques—including machine learning, deep learning, and robotics—have improved crop performance across different agricultural domains, such as crop prediction, classification, and disease management.

### 5.2 Resource Efficiency

AI helps farmers use water, fertilizers, and pesticides only when and where needed. For example, smart irrigation systems avoid overwatering, which conserves water and prevents nutrient leaching. Likewise, AI-assisted spraying drones can treat only the affected areas of a field, reducing chemical usage.

The integration of AI with big data platforms supports real-time decision-making and reduces waste, leading to more sustainable farming practices.

### 5.3 Labor and Cost Savings

Labor shortages are a major issue in agriculture, especially in remote and aging farming communities. AI technologies such as automated tractors, robotic weeders, and intelligent harvesters reduce the need for human labor. This not only lowers operational costs but also makes farming more accessible to those with limited workforce availability.

### 5.4 Enhanced Food Safety and Quality

AI-powered systems like electronic noses (e-noses) and electronic tongues (e-tongues) are being used to monitor food freshness, taste, and contamination levels during and after harvest. These systems mimic human senses and use pattern recognition to evaluate food quality quickly and accurately.

### 5.5 Climate Resilience and Environmental Protection

AI supports environmentally responsible farming. By helping farmers adapt to changing weather patterns, it enhances climate resilience. For instance, predictive models can anticipate droughts, pests, or flooding, allowing farmers to take early action.

Moreover, reducing excessive input use (like fertilizers and pesticides) lowers greenhouse gas emissions and protects biodiversity, contributing to global climate goals.

## 6. Challenges and Barriers to AI Adoption in Agriculture

Despite its potential, several barriers limit the widespread use of AI in agriculture. These challenges are technical, economic, and social in nature.

### 6.1 High Initial Cost and Infrastructure Requirements

AI-based systems require access to high-quality equipment, such as sensors, drones, cameras, and computing devices.

Many small-scale or low-income farmers cannot afford these tools or the infrastructure needed to support them, such as internet connectivity and electricity.

### 6.2 Lack of Technical Skills

Successful use of AI requires training in technology use and data interpretation. Many farmers, especially in rural areas, may not have the education or technical background to use these systems effectively.

As suggested by Cavazza *et al.* (2023) <sup>[2]</sup>, collaboration between researchers and practitioners is essential to promote knowledge sharing, training, and development of user-friendly tools.

### 6.3 Data Privacy and Ownership

AI systems often collect large amounts of data about farms, crops, and soil conditions. Questions about who owns this data, how it is used, and whether it is secure are still unresolved in many places. Farmers may hesitate to adopt such technologies if they fear losing control over their farm information.

### 6.4 Limited Localization and Customization

Many AI tools are developed using data from specific regions. As a result, they may not perform well in different environments or under different soil and climate conditions. There is a need to customize AI tools to fit local agricultural practices and cultural contexts.

### 6.5 Ethical and Social Concerns

The use of AI may raise ethical questions about job replacement, especially in labor-intensive economies. There are also concerns about decision-making transparency and the possibility of errors in automated systems that could harm crops or animals.

Ryan *et al.* (2023) <sup>[5]</sup> recommend an interdisciplinary approach to AI development in agriculture, ensuring that ethical, economic, and social factors are considered alongside technical innovation.

## 7. Case Examples of AI in Agriculture

### 7.1 Smart Weeding in India

A team from ICAR and IIT Kharagpur developed an AI-enabled mechanical weeder that uses fuzzy logic to avoid damaging crop plants during intra-row weeding. This system adjusts its weeding tools in real time based on plant spacing and soil resistance, achieving over 65% weed removal and under 25% crop damage.

### 7.2 Agri-BIGDATA in China

Researchers in Beijing developed the Agri-BIGDATA framework to optimize nitrogen use in cereal crops. By integrating remote sensing, machine learning, and field data, they created a system that helps farmers apply the right amount of nitrogen fertilizer at the right time, reducing environmental damage and cost.

### 7.3 AI in Poultry Monitoring

In China and Kenya, AI-based computer vision systems are being used to monitor poultry health and behavior. These systems detect abnormal movement patterns or stress, helping farmers address issues before they impact productivity or welfare.

## 8. Future Trends and Opportunities

The future of AI in agriculture is promising, with emerging trends and innovations continuing to reshape the field.

### 8.1 Integration with IoT and 5G

Combining AI with the Internet of Things (IoT) and 5G connectivity will allow real-time data exchange between devices in the field. This will improve the accuracy and speed of AI decisions.

### 8.2 Vertical Farming and Urban Agriculture

AI is also supporting urban agriculture through vertical farming, where crops are grown in stacked layers under controlled conditions. AI manages lighting, temperature, and watering to maximize yield in limited spaces.

### 8.3 AI-Powered Market Forecasting

AI can be used to predict market prices, consumer demand, and global supply chain risks. This helps farmers make better decisions about what crops to grow and when to sell.

### 8.4 Sustainable and Circular Agriculture

AI will play a key role in circular agriculture, where waste is minimized, and resources are reused. Predictive models will help farmers recycle nutrients, compost organic waste, and reduce emissions.

## 9. Conclusion

Artificial Intelligence is transforming agriculture into a smarter, more efficient, and environmentally sustainable sector. By enabling data-driven decisions, AI technologies such as machine learning, robotics, computer vision, and big data analytics are solving longstanding agricultural challenges—from irrigation management to pest detection and labor shortages.

Despite its advantages, the widespread adoption of AI still faces several barriers including cost, technical expertise, and infrastructure limitations. Additionally, ethical and social concerns, such as job displacement and data privacy, require thoughtful solutions. Multidisciplinary collaboration and policy support are crucial to overcoming these barriers and ensuring that the benefits of AI reach smallholder and large-scale farmers alike.

As global food demand continues to grow, AI holds the potential to significantly boost agricultural productivity while promoting sustainability and resilience. Future research and innovation must focus on inclusive, localized, and ethically responsible AI tools that align with the real needs of farmers and the planet.

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