

# Sustania: A Digital Farming Ecosystem for Smart Agriculture

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**Abstract**—Recent challenges in agricultural water management, coupled with the increasing need for sustainable farming practices, have highlighted the critical importance of efficient irrigation systems. This paper presents Sustania, an innovative Smart Irrigation System designed for small and medium-scale farmers. The system integrates Internet of Things (IoT)-based sensors, machine learning algorithms, and a user-friendly web interface to optimize water usage in agricultural settings. Sustania offers 93.94% accuracy in irrigation predictions using soil moisture and temperature sensors and a Random Forest classifier, ensuring precise and reliable scheduling. The platform has a modular architecture with an SQLite backend and a React-based frontend, making it a cost-effective solution. Preliminary tests show remarkable water savings along with better crop yields, indicating the system's capability to change conventional agricultural methods.

**Index Terms**—precision agriculture, sustainable agriculture, irrigation system, water conservation, agricultural technology, climate change, soil health, remote sensing, environmental impact assessment, smart irrigation, machine learning, IoT integration, real-time processing

## I. INTRODUCTION

Increased competition for fresh water supplies around the world coupled with the effects of climate change requires new solutions for agricultural water management. Agriculture uses about 70 percent of the world's freshwater, though at least half of this is lost through traditional irrigation methods. Over- or under-irrigation not only wastes precious water but also harms soil health and leads to lower crop yields. On top of this, changing weather patterns make it much harder for farmers to predict how best to schedule their irrigation. Sustania is a cutting-edge smart irrigation system that tackles these fundamental challenges through the incorporation of advanced technologies like the Internet of Things (IoT), machine learning, and real-time data processing. These technologies work collaboratively to irrigate management comprehensively, automatically, and intelligently. Sustania gives farmers accurate, data-driven recommendations specifically related to the needs of their crops, soil types, and environmental conditions. The system is specially designed to cater to the needs of small and medium-scale farmers, who usually operate under the limitations of resources and access to sophisticated agricultural technologies. Therefore, Sustania is meant to be inexpensive

and accessible; thus, farmers would not need to have a considerable level of technical knowledge to use it. Through enabling real-time monitoring of soil conditions using IoT-based sensors and automating irrigation schedules with a machine-learning algorithm, Sustania eliminates all guesswork and ensures optimal water utilization. Moreover, Sustania fits within larger initiatives of sustainable agriculture by decreasing water waste significantly and improving resource management. It provides farmers with valuable information about the conditions in their fields, allowing them to make decisions that not only increase productivity but also ensure the long-term sustainability of their land. Sustania's modular and scalable design makes it a flexible solution that can adjust to the various challenges that agriculture faces in a changing climate.

## II. RELATED WORK AND MARKET ANALYSIS

There has been an avalanche of applications and platforms aimed at assisting farmers in improving irrigation management and monitoring crop development. This chapter presents significant applications and websites that offer similar features to those provided by Sustania in dialogue with strategies for sustainable and innovative agricultural practices. Thus it reinforces the distinct position of Sustania within the Kerala agri-optimization market.

### A. Related Works

a) *Netafim Smart Irrigation Solutions*: Netafim is a world leader in precision irrigation technology, developing advanced systems that optimize water use in agriculture. Their solutions incorporate innovative technologies such as automated drip irrigation and remote monitoring, which further enhances irrigation efficiency. These systems are less accessible to small and medium-scale farmers due to the high installation and operational costs involved. Besides, technical expertise is required to manage and maintain the system; this can be a barrier for farmers in regions with limited access to training and support services. While Netafim excels in water-use efficiency and crop productivity, its high cost limits its use on smaller farms that have budget constraints.

b) *RainMachine*: RainMachine offers smart irrigation controllers that operate mainly based on weather data to plan and optimize watering. The devices use forecasts from local weathers to change irrigation schedules in real-time so that watering at least coincides with conditions like rainfall and temperature instead of losing water through unwarranted watering. However, while this drastically cuts down on water waste due to overwatering, RainMachine's dependence on weather data more or less locks it into a single feature. The system does not vary soil moisture levels nor does it consider crop requirements at individual levels; the two most essential elements for managing irrigation. Consequently, it may lack precision compared with systems using real-time soil-specific data and could be less effective in some agricultural situations.

c) *FarmBot*: FarmBot is an innovative system that uses advanced robotic technology to automate the planting, watering, and maintenance of crops. It features precision hardware along with adjustable software for a totally automated farming experience. On the other hand, these are not practical for most smaller and medium-scale agricultural operations due to their complexity and high upfront cost. It demands a lot of technical knowledge about systems setup and operation; thus, most farmers are likely to shy away from it. Also, such a high-tech focus on automation may not be compatible with the financial or operational capabilities of conventional agribusinesses. While it has enormous potential to revolutionize agricultural practices, its cost and complexity today limit its use to niche markets only.

## B. Market Review

Irrigation solutions available in the market are primarily designed for large-scale agricultural operations and offer a high cost and complex interface, making them impractical for small and medium-scale farmers. Besides, these systems hardly cater to the unique challenges faced by particular regions; thus their effectiveness is limited in localized settings. Sustania offers special consideration for regional scalability and affordability. Using IoT sensors combined with machine learning, it provides accurate, data-driven irrigation recommendations specific to the local soil, crop, and climate conditions. While the user-friendly interface ensures accessibility for those not technically inclined, the cost-effective design makes it feasible even for smaller agricultural operations. It is this precision coupled with simplicity and affordability that makes Sustania a viable solution to enhance water efficiency while also encouraging sustainable farming practices.

## III. LITERATURE SURVEY

The use of technology in agriculture has resulted on a significant development is irrigation optimization where the two main technologies have been IoT and machine learning. Many research studies show how these technologies improve resource efficiency as well as sustainability in agricultural practices. Kumar and Desai (2021) discuss IoT-based systems that utilize soil moisture and temperature sensors to

save water, which is a crucial finding for the real-time data aspect of precision irrigation management [1]. On the other hand, weather-based irrigation systems are incorporated into scheduling by Ravi and Ramesh (2020), thus reducing wasted water while enhancing crop productivity [2]. Nevertheless, such systems as RainMachine rely excessively on weather data instead of soil factors that determine irrigation needs; thus, their effectiveness is limited within diverse agricultural settings. Machine learning, especially Random Forest classifiers, has proven highly effective in predicting irrigation needs based on environmental data. Models developed by Singh and Jain in 2022 are found to be highly accurate in capturing soil moisture and climatic conditions, whereas Gupta and Sharma in 2021 elaborated on the contributions of supervised learning models toward better decisions in precise agriculture. These studies emphasize the significance of machine-learning algorithms in refining irrigation schedules. Sustania takes this work further by combining Random Forest models with IoT sensors for precise real-time recommendations. Unlike FarmBot, which drastically automates but is still inaccessible to most farmers due to high costs and complexity, Sustania is an affordable and scalable solution targeted at small to medium-scale farming. Precision agriculture has been praised for its ability to conserve resources and enhance productivity. While Kumar and Mishra (2020) highlighted the significance of IoT systems in facilitating automated decision-making and enhancing resource efficiency in agriculture, Sahin and Ahmed (2022) showcased how affordable IoT systems are making precision agriculture technologies more accessible. However, such systems like Netafim Smart Irrigation Solutions, which incorporate advanced drip irrigation techniques, are too expensive for smaller farms; hence they have not been widely adopted. Sustania solves this challenge with an affordable, intuitive platform that combines IoT sensors with predictive analytics to optimize water usage in a manner accessible to resource-constrained farmers. The combination of IoT and machine learning has been widely explored for sustainable agriculture. According to Patel and Singh (2022), IoT sensors along with machine learning models enable real-time monitoring and accurate irrigation predictions [7]. In the same vein, Bansal and Mehta (2021) emphasized the use efficiency of IoT data combined with Random Forest classifiers for water optimization [8]. Although RainMachine, FarmBot, and other such systems exhibit creative solutions, these lack affordability and flexibility compared to what's needed by small-and medium-scale farms. Sustania fills this gap by offering a solution designed with local needs in mind; affordable and pragmatic for farmers in resource-constrained environments.

The literature reviewed gives evidence of the revolutionary potential of IoT and machine learning in the agricultural sector, especially concerning irrigation optimization. Thus, by filling the gaps identified within current solutions and combining knowledge from earlier research, Sustania can prove to be a viable, scalable, and efficient solution toward enhanced water management and sustainable agriculture practices. Meanwhile,

instead to focus on basic strengths, it highlights capability in providing practical insights and adjusting to varied agricultural ecosystems; hence very much significant as innovation in promoting resource-efficient as well as environmentally benign farming practices.

#### IV. SYSTEM ARCHITECTURE AND DESIGN

Sustania's architecture is designed for scalability and user-focused efficiency, integrating the following key modules:

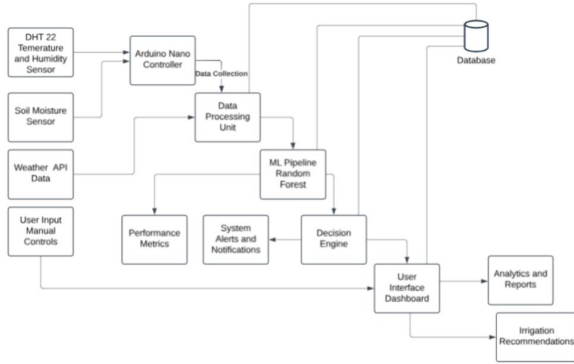


Fig. 1: System Architecture Diagram

Fig. 1 illustrates a smart irrigation system that uses sensors (for temperature, humidity, and soil moisture), weather data, and an Arduino Nano controller to collect and process data. A machine learning pipeline (Random Forest) analyzes the data to generate irrigation recommendations via a decision engine. The system integrates a user dashboard for monitoring, manual controls, and reports, while storing data in a database for analytics and performance tracking. It ensures efficient water usage with real-time alerts and insights.

##### A. IoT Sensor Integration

The system uses several sensors, specifically the DHT22 Temperature and Humidity Sensor and the Soil Moisture Sensor, to gather data about the environment. The DHT22 measures the ambient temperature and humidity, both of which are essential for determining weather conditions as well as the requirements of plants. The soil moisture sensor ensures that the water content in the soil is monitored so that irrigation decisions are based on accurate real-time data. Also, it fetches data from external APIs like Weather API Data that comprise forecasts along with environmental parameters including rainfall and temperature trends.

##### B. Arduino Nano Controller

This microcontroller serves as the central unit for collecting data from the sensors. It acts as the bridge between the physical sensors and the data processing unit. The Arduino Nano Controller gathers raw input from the environment and forwards it for further processing and analysis.

##### C. Data Processing Unit

Once data is collected, it is processed in the Data Processing Unit. This component cleans, organizes, and formats the raw data received from the sensors and the Weather API, making it suitable for analysis by the machine learning pipeline. It ensures that only relevant and accurate information is used for decision-making.

##### D. Machine Learning (ML) Pipeline

The ML pipeline employs a Random Forest algorithm to analyze the processed data. This algorithm is trained to predict irrigation needs based on historical data and real-time environmental inputs. It ensures data-driven and optimized decisions are made, improving water efficiency and reducing wastage.

##### E. Decision Engine

The decision engine interprets the output from the ML pipeline and translates it into actionable recommendations. These decisions may include specific irrigation timings, durations, or amounts based on current soil conditions, weather forecasts, and plant requirements.

##### F. User Interface Dashboard

The User Interface Dashboard is designed as an interactive platform for the user to engage with the system. The dashboard presents analysis, reports, and recommendations in a comprehensible layout. Through this feature, users can track real-time performance metrics of the system, access some historical data, and even change parameters manually if they wish to do so. The dashboard keeps the system transparent while making it flexible according to user preferences.

##### G. System Alerts and Notifications

This component keeps users informed with real-time updates about system status, alerts, and notifications. These can include warnings about abnormal conditions, recommendations for immediate action, or updates on system performance.

#### V. RESULTS

The system generates detailed performance metrics and analytics reports, offering insights into environmental conditions and the system's efficiency. Based on these insights, it provides irrigation recommendations, helping users implement precise and optimal watering strategies tailored to the specific needs of their crops or plants.

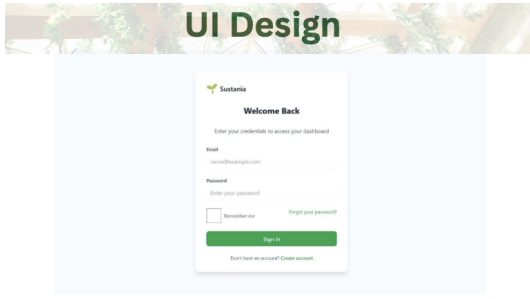


Fig. 2: Registration Page

Fig. 2 shows the website's Login page, where users need to enter their email and password to log in to the website.

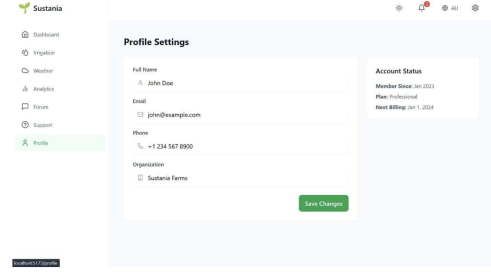


Fig. 5: Profile Settings

Fig. 5 shows the website's Profile Settings page where users can edit their profiles and add personal information.

## VI. DISCUSSION

### A. Performance Metrics

The machine learning model applied in Sustania proactively predicts irrigation needs with high accuracy. In fact, according to the evaluation metrics, the system provides an accuracy of 93.94%, which means that almost all predictions made by the model are correct. Furthermore, the precision score of 1.0000 accounts for every irrigation event predicted as needed; thus, it extenuates all assumptive water waste. The recall score is 0.8788, which foresees most instances where irrigation was required and thereby ensures minimal risk of under-irrigation. These results testify to the robustness of the Random Forest algorithm while managing diverse environmental inputs for accurate prediction. With such an accuracy level, Sustania can optimize irrigation better, hence improving resource efficiency and crop vitality.

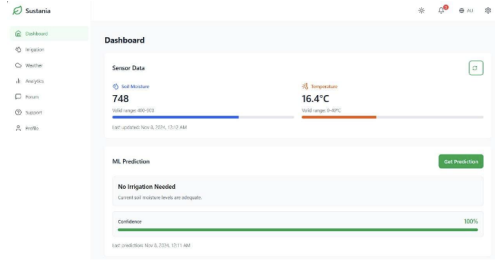


Fig. 3: Home Page

Fig. 3 shows part of the website's Farmer Dashboard page, which provides real-time readings of soil moisture and temperature along with irrigation predictions.

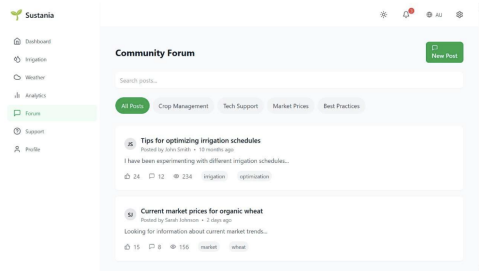


Fig. 4: Community Forum

Fig. 4 shows part of the website's Community Forum where farmers can interact with each other and specialists in the industry.

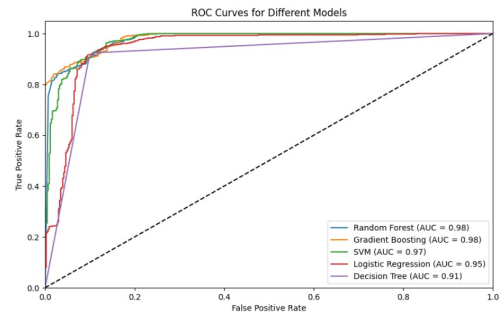


Fig. 6: ROC Curve

Fig. 6 shows the ROC curve for different ML Models. The figure clearly shows why Random Forest was selected for our system. AUC for Random Forest is one of the highest among all the models we had tested. Though Gradient Boosting too has the same AUC, Random Forest was selected because it is a little faster than Gradient Boosting.

TABLE I: Classification Report

Decision	Precision	Recall	F1-Score	Support
0	0.87	0.93	0.90	433
1	0.94	0.89	0.92	567
Accuracy			0.91	1000
Macro avg	0.91	0.91	0.91	1000
Weighted avg	0.91	0.91	0.91	1000

Table I shows the performance assessment of the trained ML Model. The output shows a machine learning model trained with 91% accuracy, highlighting "moisture" as the top feature. The best parameters were tuned, and the model was saved as irrigation\_model.joblib.

```

Command Prompt
Pitting 5 folds for each of 48 candidates, totalling 240 fits
2024-11-07 22:44:32,892 - INFO - Best parameters: {'classifier__class_weight': 'balanced_subsample',
classifier__min_samples_leaf': 0, 'classifier__min_samples_split': 5, 'classifier__n_estimators': 100}
2024-11-07 22:44:32,892 - INFO - Model training completed successfully

Classification Report:
precision    recall  f1-score   support
0           0.87    0.93    0.90       433
1           0.94    0.89    0.92       567

accuracy          0.91    0.91    0.91    1000
macro avg         0.91    0.91    0.91    1000
weighted avg      0.91    0.91    0.91    1000

2024-11-07 22:44:35,433 - INFO - Model evaluation completed successfully

Top 10 Most Important Features:
moisture: 0.2173
moisture_normalized: 0.1873
temp_normalized: 0.1624
temp: 0.1439
temp_hot: 0.0965
moisture_temp_interaction: 0.0663
moisture_low: 0.0443
moisture_medium_high: 0.0375
temp_optimal: 0.0169
moisture_medium_low: 0.0152

2024-11-07 22:44:36,008 - INFO - Model saved to irrigation_model.joblib

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Fig. 7: Performance Assessment

Fig. 7 shows the screenshot of the performance assessment of the trained ML Model.

TABLE II: Top 10 Most Important Features

Feature Name	Importance Score
moisture	0.2173
moisture_normalized	0.1873
temp_normalized	0.1624
temp	0.1439
temp_hot	0.0965
moisture_temp_interaction	0.0663
moisture_low	0.0443
moisture_medium_high	0.0375
temp_optimal	0.0169
moisture_medium_low	0.0152

Table II shows the Top 10 features. Feature importance in a Random Forest model measures how much each feature contributes to predicting the target variable. It helps identify the most influential features, enabling better model interpretation, feature selection, and performance improvement. Features with higher importance are more relevant for the model's predictions, while less important features can be removed to simplify the model.

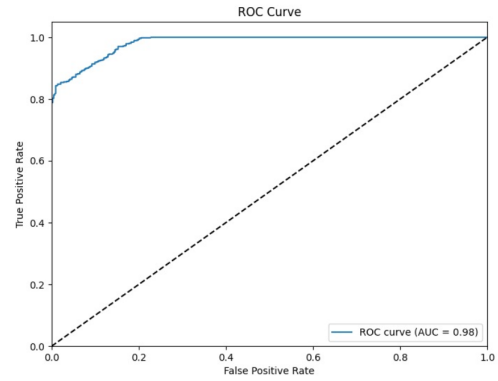


Fig. 8: ROC Curve

Fig. 8 shows the ROC curve of the Random Forest Model after training. The curve demonstrates excellent model performance with an AUC (Area Under the Curve) of 0.98, indicating the model effectively distinguishes between classes. The closer the curve is to the top-left corner, the better the performance.

### B. Scalability

Sustania is modular and flexible, which makes it easily adaptable to diverse farming conditions and scales. In terms of design, it is open to additional sensors that can be integrated into the system, like those measuring soil nutrients or weather conditions; thus, making it suitable for more complex farming operations. On small-scale farms versus large agricultural estates, Sustania's scalability ensures compatibility at various farm sizes and crop types. This gives a major advantage in the system's potential for widespread adoption as it can be tailored to the specific needs of different farming communities. Moreover, the affordability and ease of use of the solution make it accessible to small and medium-scale farmers who usually cannot afford to pay for high-end technology solutions. Because of its efficiency while operating under various agricultural conditions, it can be considered a truly versatile solution that can be applied worldwide, especially in areas where water is scarce or resources are constrained. Hence these characteristics would highlight Sustania as a sustainable and viable tool for modern agriculture to provide better water management and productivity on an immense scale.

## VII. CONCLUSION AND FUTURE SCOPE

Sustania is a solution that embodies the evolution of precision farming; it combines Internet of Things, machine learning, and an interactive interface to tackle irrigation challenges for small to medium-scale farmers. While dealing with water usage, wastage in terms of sustainability and keeping crops hydrated optimally, it proves itself a sustainable along with a cost-effective solution in various scenarios. With these two parameters at its core, Sustania connects advanced agricultural technology with resource-constrained farming communities which will soon be widespread Smart Irrigation Systems adoption. Looking Sustania then, it has a great potential of

growing and innovating. Advanced sensors could be added, for instance, those detecting soil nutrient levels, thereby enhancing the system's capability to manage crops more effectively- only not through water usage alone. Long-term irrigation needs could be forecasted using AI by analyzing data history, weather patterns, and crop cycles. This would enable the farmer to plan better and ensure sustainability in water management over time. A mobile application would further enhance the accessibility of Sustania by offering farmers instant updates, control options, and information directly on their smartphones. These additions would not only make the system more user-friendly but also help it reach agricultural communities that are quite far from the main areas.

Moreover, alliances with governmental agencies and non-governmental organizations would further sustain any impact because it could ease the wide adoption through financial support or subsidies for farmers. Partnership with research institutions in agriculture would also spice up the platform by offering models designed for specific regions that will cater to diverse area needs. All these would encourage the smart irrigation system's adoption while equipping the farmers with customized solutions and know-how to enhance productivity alongside sustainability. Sustania could standardize itself as a cornerstone in the innovation of precision agriculture by amenities making supportive networks of innovation and collaboration that drive economic growth together with environmental stewardship principles.

These future developments will reinforce Sustania's role in promoting sustainable farming practices, helping to address global challenges such as water scarcity, food security, and climate change. By continuously evolving and incorporating new technologies, Sustania has the potential to redefine irrigation management and support the long-term growth of agriculture worldwide.

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#### LIST OF ABBREVIATIONS

Application Programming Interface (API)  
 Area Under Curve (AUC)  
 Direct Current (DC)  
 Digital Humidity and Temperature Sensor (DHT22)  
 Internet of Things (IoT)  
 Joblib - Python Library for Serialization (joblib)  
 Machine Learning (ML)  
 Numerical Python (NumPy)  
 Python Data Analysis Library (Pandas)  
 Representational State Transfer Application Programming Interface (REST API)  
 Receiver Operating Characteristic (ROC)  
 Support Vector Machine (SVM)

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