



Strategic Resource Management Framework for Sustainable Agricultural Practices and Environmental Protection

Dr. M. Kalaivani^{1*}, Dr. M. Vijayakumar², Dr. M. Ganeshkumar³, Dr. S. Karunakaran⁴

^{1*} Associate Professor, Faculty of Management, SRM Institute of Science and Technology, Vadapalani campus, Chennai, Tamil Nadu, India. ORCID: <https://orcid.org/0000-0002-3593-1417>, E-mail: drkalaimba@gmail.com

² Professor, School of Management, Vels institute of Science Technology & Advanced Studies, Chennai, Tamil Nadu, India. ORCID: <https://orcid.org/0000-0002-9154-3507>, E-mail: vijai711@gmail.com

³ Assistant Professor, Faculty of Management, SRM Institute of Science and Technology, Kattankulathur Campus, Chengalpattu, Tamil Nadu, India. ORCID: <https://orcid.org/0000-0001-6828-0051>, E-mail: ganeshkm@srmist.edu.in; dr.ganeshkumarphd@gmail.com

⁴ Associate Professor, Department of Management Studies, St. Joseph's College of Engineering, OMR, Chennai, Tamil Nadu, India. ORCID: <https://orcid.org/0009-0009-2478-5352>, E-mail: karunakaranmba.s@gmail.com

Abstract

The agricultural systems across the globe are under increasing strain from the limited natural resources, unpredictable climatic patterns, and the challenge of fulfilling the twofold role of achieving food security while pursuing environmental sustainability. The present paper discusses a Strategic Resource Management Framework (SRMF), which uses data-based resources assessment along with sustainability indicators modeling and decision-support optimization to provide guidance in agricultural activities that would promote long-term environmental and economic sustainability. The framework includes four interrelated levels: the level of gathering field data, the level of assessing resources and computing indicators, the level of strategic optimization and decision support, and the level of implementation and continuous feedback monitoring. Unlike the traditional approaches which treat agricultural inputs separately, the framework presented in this paper combines soil, water, energy, and biodiversity indicators in a single analytical framework, which would allow conducting trade off-analysis in a coordinated manner. The framework is presented along with the implications of its application, among which are higher resource efficiency, reduced negative environmental impact, and greater resilience to climate-related stress.

Keywords:

Sustainable agriculture, strategic resource management, environmental protection, resource-use efficiency, decision support framework, agricultural sustainability indicators.

Available online: 01/09/2026

Introduction

Agriculture is a major backdrop of global food security as well as an important user of natural resources like land, freshwater, and soil nutrients and its performance is linked to the general problem of environment protection now (Dias et al., 2022). It is understood that the growth of the population and land cultivation leads to a lot of pressure on the limited natural resources which raises the question of the sustainability of traditional agricultural systems (Yang & Solangi, 2024). Farmers have to work within the framework of reducing economic possibilities which makes the strategic use of resources vital for competitiveness in these conditions (Aydin & Degirmenci, 2026).

In the past, resource management in agriculture was done separately where different aspects such as irrigation, fertilization, energy consumption and land use were handled in isolation from one another (Hassan et al., 2026). This separated method resulted in inefficient practices such as excessive groundwater extraction, overuse of agrochemicals and deterioration of environmental conditions (Seraj et al., 2026). Technological advancements in data analytics have made resource management possible through an integrated approach (Thooyamani et al., 2014).

Increasingly, agricultural policy makers are keen on behaviour that takes into account the previous aspects together with the processes and effectiveness of resource conservation and climate resilience and areas adopting resource management measures tend to report significant improvements, which have an improvement regarding agricultural productivity and lowered environment-related risks (Kaklauskas et al., 2026; Wongnaa et al., 2026). However, there are still many models that concentrate solely on one aspect of resource utilization, thus not being able to account for the interconnections that affect sustainability of the entire system (Feng et al., 2025). Thus, there is a need for a model that is able to combine several streams of resources to form a coherent framework that combines various resource utilization parameters (Al-Salahi et al., 2025).

Key Contributions

This paper contributes to the existing body of knowledge in the following ways

- It offers an integrated framework of strategic resource management (SRMF) taking into account soil, water, energy and biodiversity factors in a single analytical framework.
- The framework has got a four-layer architecture in data acquisition, resource evaluation, strategic optimization and implementation-feedback tracking applications for continuous improvement in resource use efficiency.
- The research sets out a set of indicators of sustainability making it possible to compare different kinds of resources instead of evaluating each of them separately.
- It discusses the future performance of the framework and its usefulness to policymakers, agribusiness specialists and scholars.

Paper Organization

The rest of the paper is structured as follows. Section 2 briefly looks at existing literature on natural resource management, environmental protection, and sustainable agriculture. The proposed framework for Strategic Resource Management is discussed in section 3 along with the structure and layers of this framework. Expected outcomes from this framework are described in section 4. The section 5 provides conclusions and suggests directions for further research.

Literature Review

Studies of small and medium-sized agricultural enterprises prove that the relationship between the presence of natural resources and the performance of businesses is conditioned by entrepreneurial orientation and environmental sustainability orientation, which suggests that the availability of resources does not guarantee success (Dias et al., 2022). Experiments on automated and virtual instrumentation systems in agriculture show how the use of sensors allows for reducing the need for manual interference (Thooyamani et al., 2014).

At the macroeconomic level, natural resources management, environmental management and agricultural economics are highly interconnected and the quality of resource management is crucial for sustainable development results (Yang & Solangi, 2024). The case of automation of farms demonstrates how the

integration of sensor systems and sustainability goals can help to manage resources in a more stable way (Al-Salahi et al., 2025). Other economic-environmental analysis questions the role of taxation, resource depletion policies, renewable energy usage and globalization on sustainable environmental results (Aydin & Degirmenci, 2026).

Deep learning-based diagnostic methods used for crop monitoring illustrate the application of advanced methods to support sustainability-focused agricultural systems (Dhanya & Mythili, 2026). The research focusing on environment and policy-related aspects of mineral resource management in relation to green growth demonstrates that the use of prioritization frameworks can achieve balance between economics and ecology with important implications for agriculture (Hassan et al., 2026). The forecast techniques based on genomics used for the agricultural biotechnology market demonstrate increasing interest in using scientific and economic models for sustainable resource management (Ajitha et al., 2025).

Water-energy-food-environment nexus research highlights several factors contributing to the sustainability of agricultural development. This underscores the idea that resource types are interdependent and not independent of each other (Seraj et al., 2026). The research done in circular bioeconomy practice along with smart agriculture encourages the use of resource recovery as well as technology within rural development initiatives (Karthika, 2025).

The adaptive policy analysis approach to sustainable agriculture is characterized by contextual and timely updating resource governance processes (Kaklauskas et al., 2026). Mitigation and adaptation approaches to climate-smart agriculture show that the process of resource management should always be responsive to the changing climate conditions (Awais et al., 2026), whereas the case study of the adoption of integrated soil fertility management reveals welfare gains for smallholder farmers (Wongnaa et al., 2026).

It has been suggested that sustainable farm systems that simultaneously achieve soil health, water use efficiency and crop productivity can serve as a strategy towards climate adaptation, similarly to the approach taken in this paper (Keshyagol et al., 2026). Studies on resource management in agriculture that include climate mitigation in terms of carbon emissions demonstrate the necessity of accounting for the outcomes of climate mitigation in addition to productivity when constructing resource frameworks (Feng et al., 2025). Thus, there is increasing attention in the scientific community towards resource-based frameworks of sustainability, but most research papers focus on specific aspects of resources only.

Methodology

In this part, the Strategic Resource Management Framework (SRMF) is discussed as an idea developed to facilitate integrated decision-making based on scientific analysis and coordination of decisions in different categories of agricultural resources management. SRMF is represented as a feedback loop of five stages of the process from gathering field data to implementation and monitoring.

Framework Architecture

Figure 1 below represents the architecture of the proposed framework. The field data collected undergoes processing, assessment, and computation of indicators, followed by optimization and then implementation and monitoring. The implementation stage is continually linked to the data collection stage by a continuous feedback loop, which enables the framework to adjust the resources' strategies according to changes in field conditions and outcomes.

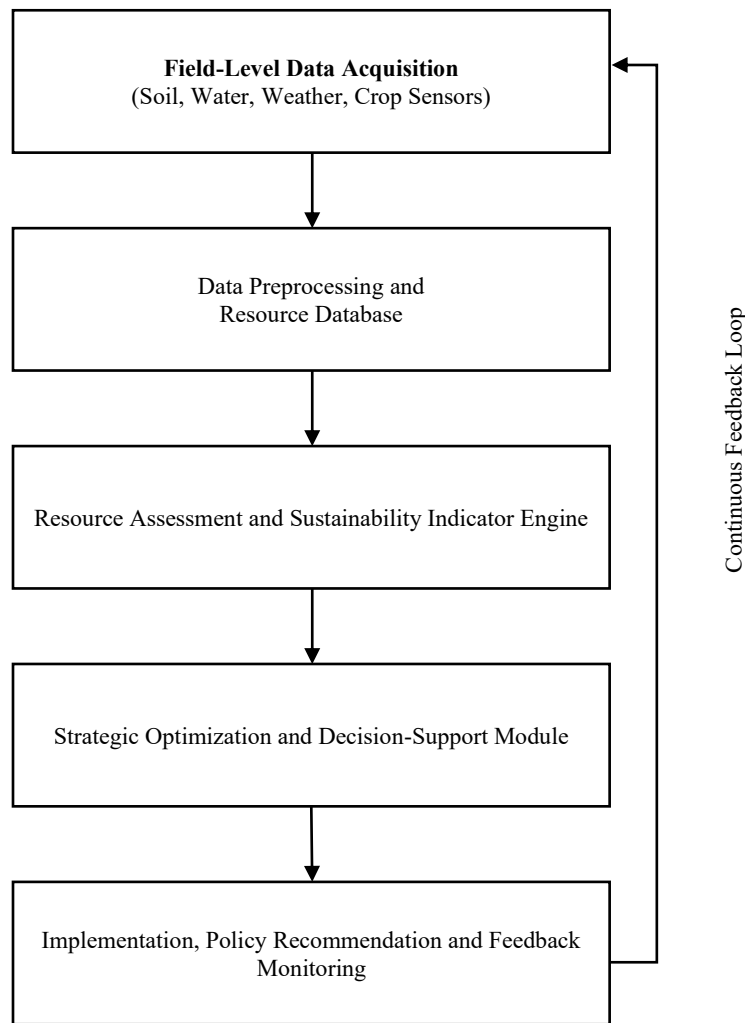


Figure 1. Architecture of the proposed strategic resource management framework

Layer Descriptions

- **Data Acquisition at Field Level:** This level collects primary data via soil sensors, weather stations, irrigation meters, and crop condition data such as soil moisture, nutrient concentration, temperature, rainfall, energy usage, and crop development stage.
- **Data Cleaning, Normalization and Resource Database:** The raw observations get cleaned and normalized, creating a unified resource database after eliminating sensor noise and bringing heterogeneous streams of information into the same temporal and spatial framework.
- **Sustainability Indicators Computation Module:** This module calculates water-use efficiency, soil health index, energy intensity, carbon footprint, and a synthetic biodiversity indicator, all in a normalized form to allow category comparison.
- **Multi-Criteria Trade-Off Analysis for Strategy Design:** A multi-objective optimization problem gets solved in this module to create a list of potential allocation strategies considering productivity, cost and environmental performance.
- **Implementation and Policy Recommendations:** Implementation of the selected strategies takes place and results go back to the data collection stage as feedback for further recommendations.

Sustainability Indicators

Five primary indices form the basis for the resource assessment engine: water use efficiency (productivity relative to volume of water used), soil index (amount of organic matter, nutritional balance, and quality), energy index (energy used relative to volume of output), carbon footprint (greenhouse gases emitted in the production process), and biodiversity index (agricultural diversity and quality of habitat). The indices are integrated into the optimization engine through the use of customizable weights, making the framework adaptable to regional concerns such as water shortage, soil rejuvenation, or emission reduction.

Decision-Support Optimization Approach

The decision support component works on a multi-objective optimization approach in which the various resource allocation strategies proposed is evaluated based on the sustainability indices discussed above. Resource allocation strategies that dominate other strategies in all the objectives is included in the feasible trade-off set. The best strategy from the trade-off set is chosen in accordance with the priorities set by the stakeholders.

Results and Discussion

Since SRMF is offered in the conceptual and architectural form, this chapter describes how the performance of SRMF should appear as per the structure of the framework's indicator system and the optimization mechanism employed, but does not describe the outcomes that result from implementing SRMF in the field. The framework should enhance resource use efficiency mainly due to the joint assessment of indicators – as water, soil, energy and biodiversity indicators are assessed simultaneously rather than separately, decisions made do not sacrifice one indicator for another.

Table 1. Illustrative comparison of conventional and SRMF-guided resource management practices

Sustainability Indicator	Conventional Practice	SRMF-Guided Practice	Expected Directional Change
Water-use efficiency	Uniform, schedule-based irrigation	Sensor-informed, demand-based irrigation	Improvement
Soil health index	Fixed fertilization regardless of soil status	Indicator-driven, targeted nutrient application	Improvement
Energy intensity	Manually scheduled equipment operation	Optimized scheduling based on demand indicators	Reduction
Carbon footprint	Emissions not explicitly tracked	Explicit tracking and trade-off weighting	Reduction
Biodiversity indicator	Limited consideration in planning	Incorporated as an explicit optimization objective	Improvement

As shown in table 1, the integrated indicator system of the SRMF is anticipated to move agricultural resource decisions away from schedule-dependent processes and towards an indicator-based system for all five sustainability domains, minimizing the possibility of contradictory advice that could result from the use of single-resource decision-making tools.

However, from the standpoint of policy, the fact that carbon footprint and biodiversity indicators have been explicitly included as part of the objective function means that the framework can incorporate environmental protection objectives in its own right rather than being a constraint on a productivity-oriented approach where the SRMF serves as a link between the two domains.

Feedback-based design also enables a resilient process over time because the outcomes from the implementation process are constantly being fed back into the data acquisition process, making the framework able to adjust its recommendation based on changes in climatic situations or prices and resources. The effectiveness of the model, however, would be determined by elements outside of this conceptual process, such as how wide the sensor network is and whether stakeholders act on the recommendations.

Conclusion

In this paper, the idea of Strategic Resource Management Framework (SRMF) is introduced that combines field-level data collection, resource evaluation, strategic optimization and implementation-monitoring feedback into one single unified approach to sustainable agricultural resource planning. Contrary to traditional methods where water, soil, energy, and biodiversity issues are addressed separately, the suggested SRMF combines these four dimensions within one unified process of assessment and optimization so that trade-offs could be managed effectively and conflicting resource management decisions would not happen. Moreover, with the help of feedback system, the recommendations could be adapted when necessary, depending on changes in field circumstances, climatic conditions and changing policies.

Future research will include empirical testing through pilot testing at the field level, calibration of sustainability indicators in regard to crop type and regional specificity, as well as improvement of the optimization algorithm to account for other socioeconomic variables like market accessibility and labour availability. Expansion of the framework to include long-term climatic scenarios would also be a valuable step.

References

1. Ajitha, K., Samuel Joseph, C., Mahila Vasanthi Thangam, D., & Lakshmi Priya, K. (2025). Genomic Insight-Enabled Economic Forecasting for Sustainable Agri-Biotech Markets. *International Academic Journal of Science and Engineering*, 12(4), 211-223. <https://doi.org/10.71086/IAJSE/V12I4/IAJSE1290>
2. Al-Salahi, S. A., Alhmood, F. J., Zakariya, K. A., Al-Muhaishi, S. A., Ahmed, T. E., Al-Shammari, R. H., ... & El-Sharawy, E. E. (2025). Enbat: Automated farm monitoring and controlling system for sustainable agriculture. *Journal of Wireless Mobile Networks, Ubiquitous Computing, and Dependable Applications*, 16(3), 354-370. <https://doi.org/10.58346/JOWUA.2025.I3.020>
3. Awais, M., Wang, X., & Ashraf, M. U. (2026). Mitigation and adaptation strategies in climate-smart agriculture: A review for sustainable production. *Climate Smart Agriculture*, 100097. <https://doi.org/10.1016/j.csag.2026.100097>
4. Aydin, M., & Degirmenci, T. (2026). Reassessing the role of resource management on sustainable environment: impacts of taxation, depletion, renewable energy, and globalization. *Sustainable Development*, 34, 879-892. <https://doi.org/10.1002/sd.70219>
5. Charpe Prasanjeet Prabhakar, & Gaurav Tamrakar. (2026). Bridging Yogic Science and Sports Medicine: An Integrative Review of Evidence-Based Approaches to Health and Performance. *Journal of Yoga, Sports, and Health Sciences*, 8-16.
6. Dhanya, R., & Mythili, S. (2026). Rice Leaf Disease Diagnosis Through Deep Learning: An Inceptionv3 Approach with Spatial Attention for Sustainable Agriculture and Food Security. *Archives for Technical Sciences*, (35), 27-48. <https://doi.org/10.70102/afts.2026.1835.027>
7. Dias, C., Rodrigues, R. G., & Ferreira, J. J. (2022). Linking natural resources and performance of small agricultural businesses: Do entrepreneurial orientation and environmental sustainability orientation matter?. *Sustainable Development*, 30(4), 713-725. <https://doi.org/10.1002/sd.2268>
8. Feng, T., Liu, B., Hu, C., Wang, Z., Liu, T., & Zhou, Z. (2025). Research on sustainable agricultural resource management considering the potential of carbon emission reduction. *Sustainable Energy Technologies and Assessments*, 83, 104641. <https://doi.org/10.1016/j.seta.2025.104641>
9. Gaurav Tamrakar, & Moti Ranjan Tandi. (2026). Bio-Inspired Energy Harvesting Mechanisms for Low-Carbon Mechanical Systems: A Comprehensive Review of Designs, Materials, and Optimization Strategies. *Advances in Mechanical Engineering and Applications*, 2(1), 25-33.
10. Gichoya David, K.L. Mdodo, Rane Kuma. (2026). Evaluating Sustainable Management Strategies for Aquatic Ecosystems Using Ecological and Socio-Environmental Indicators: A Multi-Criteria Decision Analysis Approach. *Journal of Aquatic Ecology and Environmental Sustainability*, 9-16.
11. Hassan, H., Li, C., & Ahmed, S. (2026, February). Prioritizing environmental and policy factors and solutions in sustainable mineral resources management for green growth in China. In *Natural Resources Forum* (Vol. 50, No. 1, pp. 412-434). Oxford, UK: Blackwell Publishing Ltd. <https://doi.org/10.1111/1477-8947.12579>
12. Haitham M. Snousi, & Fateh A. Alej. (2026). Reshaping STEM Entrepreneurship: Women-Led Models of Inclusive Innovation. *Journal of Women, Innovation, and Technological Empowerment*, 2(1), 17-23.
13. Kaklauskas, A., Jayasinghe, G., Rajib, S., Kaklauskienė, L., Ubarte, I., Naimaviciene, J., ... & Milevicius, V. (2026). Adaptive policy analysis of sustainable and productive agriculture and its context. *Land Use Policy*, 167, 108063. <https://doi.org/10.1016/j.landusepol.2026.108063>
14. Karthika, J. (2025). Integration of Circular Bioeconomy Practices and Smart Agriculture Technologies for Sustainable Rural Development. *National Journal of Smart Agriculture and Rural Innovation*, 3(1), 8-14.
15. Keshyagol, K., Vanarotti, M., Madiwal, S., Kulkarni, P., & Bongale, A. M. (2026). Sustainable farming systems integrating soil health water use efficiency and crop productivity for climate resilience. *Discover Sustainability*, 7(1), 481. <https://doi.org/10.1007/s43621-026-02855-0>
16. Latha B. (2026). IoT-Enabled Grain Storage Monitoring Framework for Predictive Spoilage Detection and Pest Control. *Journal of Environmental Sustainability, Climate Resilience, and Agro-Ecosystems*, 9-15.
17. R. Vrba. (2026). Energy-Efficient Smart Waste Management System Using Computer Vision and IoT for Circular Cities. *Journal of Smart Infrastructure and Environmental Sustainability*, 3(2), 34-39.

18. Seraj, M., Pervar, A., Ozdeser, H., & Seraj, F. T. (2026). Determinants of Sustainable Development of Agriculture in the WEF-E-Nexus Approach in OECD-EU Member States. *Central European Economic Journal*, 13(60), 208-226. <https://doi.org/10.2478/ceej-2026-0012>
19. Thooyamani, K. P., Khanaa, V., & Udayakumar, R. (2014). Virtual instrumentation-based process of agriculture by automation. *Middle-East Journal of Scientific Research*, 20(12), 2604-2612.
20. Wongnaa, C. A., Tawiah, F. O., Ogunleye, A. S., Prah, S., Ampomah, M. A., Akweley, G. A., & Gyekye, A. (2026). Impact of adoption of Integrated Soil Fertility Management on household welfare: evidence from tomato farmers in Ghana. *Agricultural and Resource Economics Review*, 55(1), 76-103. <https://doi.org/10.1017/age.2025.10014>
21. Yang, Z., & Solangi, Y. A. (2024). Analyzing the relationship between natural resource management, environmental protection, and agricultural economics for sustainable development in China. *Journal of Cleaner Production*, 450, 141862. <https://doi.org/10.1016/j.jclepro.2024.141862>