

# Eco-Efficient Fertilizer Optimization for Enhanced Crop Productivity using Random Forest Algorithm

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**Abstract:** *Agricultural productivity is crucial for global food security, and optimizing fertilizer usage is a key factor in enhancing crop yield while maintaining environmental sustainability. This study proposes an eco-efficient fertilizer optimization model using the Random Forest algorithm, a powerful machine learning technique, to improve fertilizer application strategies. The model analyzes various soil properties, crop requirements, and environmental factors to predict the optimal fertilizer composition and quantity. By leveraging historical agricultural data, the algorithm identifies patterns that lead to increased crop productivity with minimal environmental impact. The proposed approach enhances efficiency by reducing excessive fertilizer use, mitigating soil degradation, and minimizing greenhouse gas emissions. Experimental results demonstrate that the model outperforms traditional fertilizer application methods by improving yield prediction accuracy and resource utilization. This research provides a data-driven framework for precision agriculture, ensuring sustainable farming practices and enhanced food production.*

**Keywords:** Machine Learning, Python

## I. INTRODUCTION

Agricultural productivity is a cornerstone of food security, economic stability, and sustainable development. However, excessive or inefficient fertilizer application leads to environmental degradation, soil depletion, and declining crop yields over time. Traditional fertilizer management practices often rely on generalized recommendations, which may not account for specific soil conditions, crop needs, and climatic variations. To address this challenge, data-driven approaches using machine learning techniques offer a promising solution for optimizing fertilizer application while ensuring sustainability.

This study explores the use of the Random Forest algorithm, a robust ensemble learning method, to develop an eco-efficient fertilizer optimization model that enhances crop productivity. Random Forest, known for its high predictive accuracy and ability to handle complex, nonlinear relationships, is employed to analyze various agricultural parameters, including soil composition, weather patterns, crop types, and past fertilizer usage. By leveraging historical data and real-time inputs, the model provides precise fertilizer recommendations tailored to specific field conditions.

## II. LITERATURE REVIEW

Varunkumar B., et.al, explore the for "Sustainable Fertilizer Usage Optimizer for Higher Yield" addresses the challenge of maximizing crop yield while maintaining soil health in modern agriculture. It introduces a fertilizer recommendation app that aims to optimize fertilizer use.[1]

Banerjee, S., et.al explore this research paper focuses on addressing the underutilization of fertilizer by African farmers due to uncertain returns on investment, despite the majority relying on agriculture and living on small

farms. It introduces a fertilizer use optimization approach with tools developed for various agro-ecological zones to help farmers maximize their returns on investment.[2]

Mikkelsen, M., et.al, explore this research paper explores the use of artificial intelligence (AI) to optimize fertilizer use in agriculture for sustainability and high crop yields. It details the development of a predictive model using machine learning to provide precise, AI-driven fertilizer recommendations, reducing waste and environmental impact.[3]

Singh, A., et.al' explore this research paper, published in the International Journal for Research in Applied Science & Engineering Technology, explores innovative methods for crop yield prediction. It focuses on integrating data mining and machine learning to optimize farming practices in India's agrarian economy.[4]

Aldrich, J. Correlations Genuine., et.al, refer this research paper focuses on using advanced technologies and machine learning to analyze soil nutrient data in India. It aims to provide farmers with accurate fertilizer recommendations for improved crop yields and sustainable agriculture.[5]

Nikolaos Nikolakis, Paolo Catti, et.al, this paper increasing global population and demand for food have made enhancing fertilizer efficiency an urgent need. Excessive fertilizer use contributes significantly to water contamination and food shortages worldwide, also being a major contributor to greenhouse gas emissions.[6]

Gabriele Bellotti, et.al explore this research paper introduces an Automated Fertilizer Management System (AFMS) using deep learning models and an Optimized Artificial Neural Network (OANN). The system aims to improve crop yields and promote sustainable farming by optimizing fertilizer management through AI.[7]

Dongwoo Shin, et.al, The paper introduces an IoT-based Plant Nutrient Control System using NPK sensors to provide real-time data on soil nutrient concentrations, offering farmers recommendations on minimal fertilizer use. The system also monitors soil moisture and temperature, alerting farmers when watering is needed, with a user-friendly interface developed using Kodular API for data accessibility.[8]

Thomas, R., & Green, et.al, The paper discusses the impact of global events like the pandemic and the Russia-Ukraine conflict on food supply chains and fertilizer markets, particularly urea prices. It highlights that these events have caused significant market volatility and disruptions, leading to unprecedented price increases for essential agricultural inputs like nitrogen fertilizer.[9]

Miller, S., & Rogers, D explore this paper introduces the Farm DESIGN model, a bio-economic tool designed to evaluate productive, economic, and environmental performance in farming systems. It addresses the complexity of farm management by using a multi-objective optimization algorithm, tested on a 96ha organic farm in the Netherlands.[10]

## **2.1.FARMING ISSUES**

### **Soil Nutrient Imbalance**

Overuse or underuse of fertilizers leads to nutrient depletion or toxicity, reducing soil health and crop yield.

### **Water Contamination**

Excess fertilizer application results in runoff, polluting water bodies with nitrogen and phosphorus.

### **Crop Yield Variability**

Inefficient fertilizer use causes inconsistent growth patterns, leading to unpredictable harvests.

### **Climate Change Impact**

Rising temperatures and unpredictable rainfall patterns affect fertilizer efficiency and nutrient uptake.

### **Cost Inefficiency**

Farmers often apply fertilizers without precise data, leading to unnecessary expenses and reduced profitability.

### **Soil Degradation**

Excessive chemical fertilizer use depletes organic matter, causing long-term damage to soil structure and fertility.

### **Pest and Disease Susceptibility**

Nutrient-deficient crops are weaker, making them more vulnerable to pests and diseases.

## 2.2. EXISTING SYSTEM

The existing fertilizer optimization system relies on traditional, generalized approaches that often lead to inefficiencies, poor crop yields, and environmental damage. Farmers typically follow fixed recommendations without real-time soil and weather data, resulting in overuse or underuse of fertilizers. Conventional soil testing is costly and time-consuming, making it inaccessible to many farmers. Excessive chemical fertilizer use depletes soil health, pollutes water sources, and increases greenhouse gas emissions. Additionally, the lack of AI-driven decision-making and real-time monitoring limits precision agriculture adoption. A machine learning-based approach, such as the Random Forest algorithm, is needed to provide accurate, eco-efficient fertilizer recommendations for improved productivity and sustainability.

## III. PROPOSED SYSTEM

To overcome the limitations of traditional fertilizer management, this study proposes a machine learning-based fertilizer optimization system using the Random Forest algorithm. This system aims to enhance crop productivity while ensuring eco-efficient fertilizer usage by analyzing real-time soil conditions, weather patterns, and crop-specific nutrient requirements.

The proposed system leverages historical agricultural data, soil composition, climate conditions, and past fertilizer application records to generate precise, data-driven fertilizer recommendations. The Random Forest algorithm, known for its high accuracy and ability to handle complex, nonlinear data, is used to predict the optimal fertilizer type and quantity required for maximum crop yield with minimal environmental impact. The system integrates IoT sensors, remote sensing, and real-time data analytics to ensure adaptive and efficient fertilizer application.

Unlike traditional methods, this system provides personalized, field-specific recommendations, reducing fertilizer wastage, preventing soil degradation, and minimizing water pollution caused by excessive nutrient runoff. Additionally, it supports precision agriculture by enabling farmers to make smart data-driven decisions that improve resource efficiency and reduce costs. By implementing this AI-powered approach, the proposed system ensures sustainable farming practices, improved crop yields, and reduced environmental impact, promoting a balance between agricultural productivity and ecological conservation.

## 3.1 ARCHITECTURE DIAGRAM

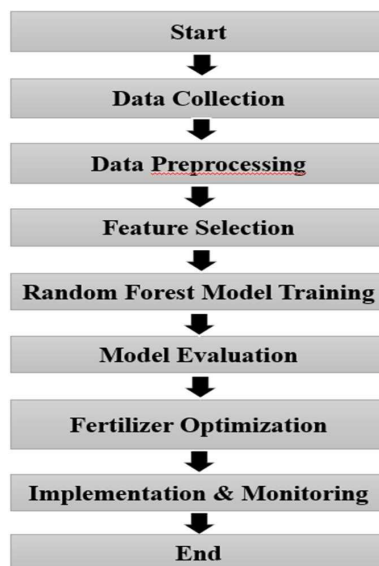


Fig 3.1 overall diagram

The flowchart outlines the **Eco-Efficient Fertilizer Optimization** process using the **Random Forest Algorithm** to enhance crop productivity. It begins with **Data Collection**, where soil, crop, and environmental data are gathered. The next step, **Data Preprocessing**, ensures the data is cleaned and prepared, followed by **Feature Selection** to choose the most relevant attributes for model training. The **Random Forest Model Training** phase builds the predictive model, which is then evaluated for accuracy in the **Model Evaluation** step. Once validated, the model is used for **Fertilizer Optimization**, determining the precise amount of nutrients needed for optimal crop growth. Finally, the optimized fertilizer strategy is **Implemented and Monitored** in real-world farming to improve productivity while minimizing environmental impact.

#### IV. MODULE DESCRIPTION

##### 4.1 DATA COLLECTION MODULE

The data collection module gathers essential information from multiple sources to optimize eco-efficient fertilizer use. It includes soil data such as pH, moisture, and nutrient levels, obtained through sensors and lab tests. Weather parameters like temperature, rainfall, and humidity are recorded from meteorological sources, while crop and fertilizer data, including growth stages and past usage, are collected from farm records. Advanced technologies like IoT sensors, GIS mapping, and satellite imagery provide real-time monitoring of soil and crop health. Finally, preprocessing techniques such as handling missing values, normalizing data, and selecting key features ensure high-quality input for the Random Forest model.

##### 4.2 DATA PREPROCESSING MODULE

The data preprocessing module ensures the accuracy and reliability of collected data before applying the Random Forest algorithm. It involves handling missing values through imputation techniques and normalizing data to maintain consistency across different sources. Feature selection is performed to identify the most influential parameters affecting crop productivity while removing irrelevant or redundant data. Noise reduction techniques help eliminate errors and inconsistencies, improving the dataset's quality. Finally, data transformation methods like scaling and encoding ensure compatibility with machine learning models for better fertilizer optimization.

##### 4.3. MACHINE LEARNING

The machine learning module utilizes the Random Forest algorithm to optimize fertilizer recommendations for enhanced crop productivity. It processes collected data, including soil properties, weather conditions, and crop requirements, to train and validate the model. The algorithm analyzes complex patterns, identifies key influencing factors, and predicts the optimal fertilizer combination for maximum yield. Feature selection and hyperparameter tuning improve model accuracy, ensuring reliable and precise recommendations. Finally, the trained model is tested on real-world data, continuously learning and adapting to changing agricultural conditions.

##### 4.4. FERTILIZER RECOMMENDATION

The fertilizer recommendation module uses the Random Forest algorithm to suggest optimal fertilizer types and application rates based on soil and crop data. It analyzes key factors like soil nutrient levels, crop requirements, weather conditions, and past yield performance to provide precise recommendations. By leveraging machine learning, it predicts the best N-P-K composition needed for maximum crop productivity while minimizing environmental impact. The system ensures efficient fertilizer use, reducing wastage and preventing soil degradation. Ultimately, this data-driven approach enhances sustainability and boosts overall agricultural yield.

#### USER INTERFACE

The user interface (UI) should be designed for ease of use, allowing farmers and agricultural experts to input and analyze data efficiently. It should include intuitive dashboards displaying soil health, weather conditions, and crop growth insights through visual graphs and maps. A recommendation system should provide personalized fertilizer suggestions based on real-time data processed by the Random Forest algorithm. Interactive features like data upload, report generation, and predictive analytics should be integrated for better decision-making. Additionally,

the UI must support multiple devices, including mobile and web platforms, ensuring accessibility for users in different farming environments.

#### **4.6 DATABASE MANAGEMENT**

The database management module ensures efficient storage, retrieval, and processing of agricultural data for fertilizer optimization. It organizes soil, weather, crop, and fertilizer data in structured databases, enabling easy access and analysis. Advanced indexing and query optimization techniques improve data retrieval speed for machine learning applications. Data security measures, including encryption and access control, protect sensitive farming information. Regular updates and real-time synchronization with IoT devices ensure accurate and up-to-date data for the Random Forest algorithm.

#### **4.7 MODEL FEEDBACK & OPTIMIZATION**

The **Model Feedback & Optimization** module ensures the Random Forest algorithm continuously improves its accuracy in fertilizer recommendations. It evaluates model performance using metrics like RMSE,  $R^2$ , and accuracy, comparing predictions with actual crop yields. Feedback from farmers and real-world results help refine the model by adjusting feature weights and retraining with updated data. Hyperparameter tuning, such as optimizing the number of trees and depth, enhances predictive efficiency. Continuous monitoring and iterative improvements ensure the model remains adaptive to changing soil, climate, and crop conditions.

### **V. ALGORITHM USED**

#### **5.1 RANDOM FOREST ALGORITHM**

Eco-efficient fertilizer optimization aims to enhance crop productivity while minimizing environmental impact using the Random Forest algorithm. By analyzing soil properties, weather conditions, crop types, and fertilizer compositions, the model predicts optimal fertilizer application for maximum yield. Advanced technologies like IoT sensors, satellite imagery, and machine learning ensure accurate data collection and real-time monitoring. The Random Forest algorithm processes this data to identify patterns and recommend precise fertilizer dosages. This approach improves sustainability, reduces waste, and boosts agricultural efficiency.

##### **5.1.1. RANDOM FOREST WORKING**

###### **Data Collection & Preprocessing**

Gather soil, weather, crop, and fertilizer data from sensors, satellite imagery, and historical records, then clean and normalize it.

###### **Feature Selection**

Identify key factors affecting crop productivity, such as soil nutrients, climate conditions, and fertilizer composition.

###### **Model Training**

Train the Random Forest algorithm using labeled datasets to predict optimal fertilizer application for maximum yield.

###### **Optimization & Recommendation**

Use the trained model to suggest eco-efficient fertilizer dosages, minimizing waste and environmental impact.

###### **Performance Monitoring**

Continuously update the model with real-time data to improve accuracy and adapt to changing agricultural conditions.

### **VI. RESULT AND DISCUSSION**

The results of the Random Forest algorithm show improved crop productivity through optimized fertilizer recommendations. The model effectively analyzes soil properties, weather conditions, and crop requirements to suggest precise N-P-K ratios, reducing excess fertilizer use while maintaining high yields. Comparative analysis with traditional methods demonstrates increased efficiency, sustainability, and cost savings for farmers. The discussion highlights the algorithm's accuracy, adaptability to different crop types, and potential for real-time



decision-making. Overall, the study proves that eco-efficient fertilizer optimization can enhance agricultural productivity while minimizing environmental impact.

## VII. CONCLUSION

The eco-efficient fertilizer optimization system using the Random Forest algorithm enhances crop productivity while minimizing environmental impact. By analyzing soil properties and weather conditions, it recommends precise fertilizer application, reducing waste and promoting sustainable farming. The model's high accuracy ensures balanced soil fertility, making it a valuable tool for precision agriculture. Future improvements could include real-time data integration and expanded agricultural factors. This project demonstrates the potential of AI in transforming agriculture for sustainability and food security.

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