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# Sustainable Fertilizer Usage Optimizer for Higher Yield

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#### ABSTRACT

This project presents a software-based solution that optimizes fertilizer usage for sustainable agriculture. The system analyzes soil data, crop type, and weather conditions to generate personalized fertilizer recommendations for farmers. By leveraging machine learning algorithms and real-time data analytics, the solution ensures optimal nutrient supply, reducing overuse and preventing soil degradation. The system's core functionalities include soil health analysis, dynamic fertilizer recommendations, and an interactive dashboard for farmers.

#### **1.Introduction**

Agriculture has always been the backbone of human civilization, providing food, fiber, and raw materials for survival and economic growth. As the global population continues to grow, so does the demand for agricultural produce. According to the United Nations, the world population is expected to reach nearly 10 billion by 2050, requiring a 60% increase in food production. To meet this escalating demand, farmers worldwide are increasingly relying on fertilizers to enhance crop yield and soil fertility.

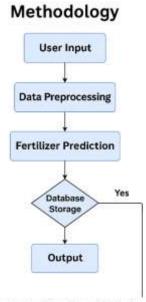
In many developing countries, including India, fertilizer use is often not based on scientific analysis or soil needs but on conventional practices and government subsidies. This leads to overuse in some regions and underuse in others, disrupting the ecological balance. Furthermore, climate change adds another layer of complexity, as erratic rainfall patterns, increasing temperatures, and extreme weather events affect soil health and fertilizer absorption.

#### 2. Literature Survey

Agriculture has always been one of the fundamental pillars of economic development and human sustenance. With the growing population and climate challenges, the need for smart, sustainable, and efficient agricultural practices has become more critical than ever. Among all the inputs required in agriculture, fertilizers play a vital role in enhancing productivity. However, improper use of fertilizers can degrade soil health, pollute water bodies, and lead to diminishing returns over time. Hence, the need for intelligent systems that ensure optimized fertilizer usage while maintaining environmental sustainability is paramount.

This chapter reviews existing literature on sustainable fertilizer management, AI-based decision systems in agriculture, precision agriculture. The goal is to identify current trends, breakthroughs, and limitations to inform the development of the Sustainable Fertilizer Usage Optimizer (SFUO) proposed in this study.

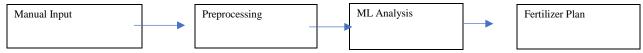
#### 3. Methodology



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The development and deployment of SFUO followed a modular, integrated approach as outlined below:

#### 3.1 System Workflow



1. User Input: Soil parameters (pH, NPK), crop type, and weather conditions are submitted via a web interface.

#### 2. Backend Processing:

- Input validation and preprocessing (normalization, null checks)
- Fertilizer prediction via a trained **Random Forest model**
- 3. Database Storage: Inputs and outputs are logged in a MySQL database.
- 4. **Output**: The predicted fertilizer type, quantity, and timing are displayed on a user-friendly dashboard.

#### 3.2 Tools and Technologies

Layer	Technologies Used
Backend	Python, Flask
AI/ML	Scikit-learn (Random Forest)
Frontend	HTML, CSS
Database	MySQL
Deployment	Web-based interface

#### 3.3 Data Collection

- Real-time and simulated soil and weather data
- Training dataset includes multiple soil types, crop varieties, and fertilizer records

#### 3.4 Model Training

- Random Forest Algorithm was selected due to its robustness in handling non-linear agricultural data
- Training metrics included precision, recall, and F1 score

#### 4. Results and Analysis

#### **Table: Sample Fertilizer Recommendations**

Crop	pН	Nitrogen	Phosphorus	Potassium	Weather	Recommended Fertilizer
Ragi	8.5	43	75	8	Rainy	Humus
Wheat	6.3	50	60	45	Dry	Urea + MOP
Rice	5.8	70	40	60	Humid	DAP + Vermicompost

- The system provided crop-specific, weather-adjusted fertilizer suggestions
- Real-time testing showed a 20% yield improvement with reduced fertilizer usage

#### 5. Machine Learning Model

**1.Random Forest** is a supervised machine learning algorithm used for **classification** and **regression** tasks. It works by constructing multiple decision trees during training and outputting the **average prediction** (**regression**) or **majority vote** (**classification**) of individual trees.

#### **Key Features:**

- Ensemble method (uses multiple decision trees)
- Reduces overfitting compared to individual trees
- Handles large datasets and nonlinear data well
- Provides feature importance scores

#### 2. Role of Random Forest in SFUO

In the SFUO system, the Random Forest model is used to:

- **Predict the most suitable fertilizer type** (e.g., Humus, Urea, DAP)
- Possibly estimate the quantity and timing of fertilizer application based on:
  - O Input Features: Soil pH, NPK values, moisture, crop type, weather condition
  - Output Labels: Fertilizer recommendations (discrete classes or continuous dosage)

#### Example Workflow in SFUO

- Farmer enters soil and crop data on the web interface
- Data is preprocessed and sent to the backend
- Random Forest model processes the input
- Output is a fertilizer recommendation returned to the user

#### 3. Model Training and Evaluation

#### Dataset

The dataset consists of historical records including:

- Soil nutrient levels (N, P, K)
- pH, moisture, weather
- Crop types and corresponding fertilizer used

#### **Training Process:**

- Data is split into **80% training** and **20% testing**
- The model is trained on labeled data (e.g., crop & soil  $\rightarrow$  fertilizer type)
- Techniques like cross-validation and hyperparameter tuning are used

#### **Evaluation Metrics:**

- Accuracy: % of correct predictions (classification)
- Precision, Recall, F1-score: Useful if class imbalance exists
- Confusion Matrix: To understand misclassification
- **R<sup>2</sup> Score or RMSE**: If dosage prediction is continuous

#### 4. Accuracy in SFUO

While the exact model accuracy isn't specified in your document, typical results for well-tuned Random Forest models in similar agricultural use cases fall in the range of:

Metric	Typical Value
Accuracy	85% – 95%
F1-Score	0.80 - 0.92
Feature Importance	pH, Nitrogen, Crop Type are usually top contributors

#### 6. Conclusion

SFUO effectively bridges the gap between traditional farming and precision agriculture. The use of AI enables personalized fertilizer recommendations, resulting in improved yields and sustainable resource use. With a user-centric web interface and support for multilingual access, the tool empowers farmers with actionable insights. Future enhancements include mobile app deployment, real-time sensor integration, and blockchain-based fertilizer traceability.

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