

## RESEARCH ARTICLE/REVIEW

# A Data-Driven Crop and Soil Management System for Sustainable Farming

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**Daniyal Ayyub Khan**

*Computer Science, MVLU College, India daniyalkhan5025@gmail.com*

**Abstract:** Farming faces growing pressure from poor crop choices, tired soils, wasteful watering, shifting weather patterns, along with missing quick access to useful guidance. Many farms suffer because they lack proper soil checks and catch pest problems too late - this throws off nutrients, cuts harvests, brings financial strain. What stands out is how badly farmers need smart ways to make decisions using solid information at the right moment. Here comes an idea: a system focused on crops and soil care that sharpens insights into land condition, matches plants better to fields, makes smarter use of resources - all fed by past records and live farm data. Looking back at how soil health has changed helps show what happens to crops when weather shifts. Because models mix ground facts, climate patterns, and past harvests, guessing yields gets better. Data tools sort out nutrient levels, rainfall records, and planting timelines in smart ways. Tests reveal smarter advice for sowing and watering - less waste shows up too. When soil and crop decisions follow clear insights, farms grow stronger, avoid losses, stay green.

**Keywords:** crop management, soil health, data analytics, smart agriculture, sustainable farming

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## 1. Introduction

Growing crops well matters for feeding people, supporting economies, leaving lives untouched by hunger. Yet millions of growers face stubborn problems harming harvests, weakening future chances. Erosion digs into fields; hard-packed earth blocks roots spreading; salt builds up where it should not be; nutrients vanish without replacement - each one chips away at what soil can do. Around a third of Earth's dirt has lost its strength already, worn down past usefulness, pushing yields lower year after year. When weather turns harsh - and it often does - the damage multiplies fast under thinning topsoil. What stands beneath our feet shapes how full plates stay tomorrow.

Farmers in places like India often depend on rain because few fields have reliable watering systems. Because of that, they must decide when to plant without up-to-date info about the weather or ground conditions. Sudden shifts in climate - rains arriving late, heat building fast - affect how crops grow, limit water availability, and push growers toward expensive ways to keep yields alive. On top of these issues sits weak storage; nearly a third to two-fifths of harvests spoil from poor transport and facilities across India. Pressure from nature mixes with outdated logistics, cutting into earnings while weakening food output - and making smarter farming methods more than just helpful.

Lately, more attention has gone toward single parts of farming challenges - like how weather shifts affect crops, what nutrients live in dirt, or when pests might show up. Still, most work misses bigger picture tools that bring together different kinds of farm information to help growers choose better paths. Take soil problems paired with unpredictable seasons - they hurt harvests, yes, but few systems blend ground details, air patterns, plant progress, and instant guidance for picking seeds, timing water, or feeding fields. Without these joined-up methods, many farmers struggle to keep pace with changing climates and tight resources, especially where support is thin. That reality hits hardest where it matters most: the field itself.

Looking into past farm data helps spot how soil traits, weather shifts, and planting routines affect harvest results. This project builds a system using field information to guide smarter choices in growing crops and caring for land. A key idea here is that combining different types of soil and climate records uncovers hidden links between what farmers put in and what they get out. With clearer insights, decisions around which plants to grow and how to use resources become more effective. Better predictions mean less waste, stronger output, improved preparation. Understanding patterns in the numbers shapes practical steps forward.

A fresh look at farming data begins with gathering an existing collection of information on soil traits - like acidity, water levels, and nutrients - alongside weather patterns and harvest results. Patterns emerge when strong number-crunching methods meet visual displays, revealing links that point toward smarter growing choices. Tools built for heavy-duty math sort through this data, checking how past earth and sky conditions shaped what grew well - or didn't. Findings aim to show clear reasons behind good or poor crops under many different natural settings. What grows depends heavily on hidden forces now brought into view. Not every finding fits neatly at first glance, yet patterns in soil traits clearly tie to how much crops produce. Weather shapes harvests more than expected, showing up again and again across seasons. When farm choices rely on solid measurements, matching plants to land gets easier. Smart adjustments follow - less waste of resources, sharper timing. Drought risks, insect threats come into view before damage starts. Leaders who shape rules take note; so do advisors working fieldside with growers. Communities depending on agriculture see a path forward, even when conditions shift unpredictably.

## 2. Literature Review: Data-Driven Systems for Sustainable Crop and Soil Management

### Literature Review: Data-Driven Systems for Sustainable Crop and Soil Management

Out here, farming began changing fast when numbers started guiding decisions instead of just habits. Tools got smarter, helping grow more without harming nature so much. With more mouths to feed every year, doing things the old way stopped making sense. Some scientists say it's not about gadgets alone - it's how farms begin thinking differently across time and space. Picture fields whispering their needs through sensors linked by invisible threads. That kind of web pulls together AI smarts and constant watchful signals from devices out in soil and sun. Real-time updates help balance water, nutrients, and timing better than before. Studies back this blend as a path that fits long-term living with land - not against it (Weraikat et al., 2024; Jaber Abdulhussaine & Aljali, 2023).

### Evolution of Soil Fertility and Nutrient Management

#### Soil Fertility and Nutrient Management Over Time

Soil health studies started by checking lab results for things like acidity, salt levels, organic material, and key nutrients. In Kolhapur, crop success was found tied more to fertile ground and water access than anything else - poor handling of plant food weakened earth over time (Ghodake, 2019). When researchers looked at different landscapes in Nepal, forests held better texture

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and chemistry compared to fields or open meadows, showing why local conditions must guide analysis (Shrestha & Kafle, 2020). Out in the Ethiopian highlands, burning soil through the "guie" method has shown clear harm - slashing both organic carbon and nitrogen levels if left unchecked. A different angle emerged though, where careful heat treatment actually helped lock in carbon while boosting access to phosphorus. This hints that old techniques might not need tossing out, just smarter oversight guided by research findings (Amare et al., 2013).

Soil science has made progress through changes in how we treat land - fixing tired ground isn't just about adding nutrients anymore. Biochar steps into the picture when erosion is a problem, cutting down dirt washed away by half and slowing water flow almost the same amount, thanks to better gaps inside soil and clumps holding firm (Nugraha et al., 2020). Think of fields in Nepal before rice grows: planting certain green plants like black gram or sunn hemp sets the stage later on; next season's wheat harvest jumps nearly 40%, with richer organic material and more nitrogen saved underground (Gautam et al., 2021). Down in Indonesia, mixing natural chicken waste with helpful microbes over years shows clear results - soybeans grow stronger because earth becomes looser, friendlier to tiny life below (Suchayono et al., 2019).

### **Iot Based Soil Monitoring And Precision Nutrient Management**

Starting with fresh data, scientists now turn old soil knowledge into live digital tools using internet-connected devices. Instead of guesswork, tiny computers like Arduino and ESP32 work alongside NPK sensors to check nutrients fast, right where crops grow (Shahab et al., 2022; Bachhav et al., 2021). Down in Kerala's rubber fields, one test showed results off by less than two percent when measuring pH and key elements - making feeding plants more accurate (Sharma & Shivandu, 2022). On another path, smartphone apps powered by artificial intelligence mix plant health signals with ground facts, leading to smarter fertilizer choices

while cutting down excess (Atalla et al., 2023).

### **Precision Water Management and Smart Irrigation**

Since farming uses lots of fresh water worldwide, smarter watering methods have drawn more attention lately. Instead of just checking ground wetness and weather like early internet-connected setups did, newer models took a different path. Those earlier versions missed out on live data sharing through online platforms or smartphone access, studies show. A shift came with tools like AgriSens, where timing for watering adjusted itself based on how crops developed over time. Remote control became possible, automation improved, results climbed past older techniques by more than one-tenth in output terms.

Improvements kept going thanks to smarter control methods. Not long after, research showed fuzzy logic controllers beat older average-based ones by saving about 61% on water without hurting soil quality (Shahab et al., 2022). Running on solar energy, combined irrigation setups turned out practical for remote areas too. These designs cut down power needs along with running expenses (Birhan & Tekalign, 2022).

### **Data Driven Farming Tips and Harvest Numbers**

Farming choices once relied on gut feeling, now shaped by smart software blending data from space and soil. Machines that help pick crops cut down mistakes people make when tallying by hand (Miao et al., 2021). Images snapped high above Earth, capturing unseen light bands, map fields and track salt buildup in dirt - useful for growing food without wrecking the land (Saha et al., 2022).

Soil nutrients, rain levels, heat patterns, and acidity shape what grows best - prediction tools now spot these links well. Not every algorithm handles this equally; one method called Random Forest often lands near 97.35 percent right calls on which crops fit where (Jaber Abdulhussaine & Aljali, 2023; Bachhav et al., 2021). When genes meet weather shifts, deeper clues emerge about plant survival, helping grow stronger types that handle pressure (Miao et al., 2021).

### **Farming forecasts built on past records and smart analysis now support steady food growth.**

Instead of waiting, farmers can spot plant troubles fast using light-sensing images and visual computing methods. One system called MangoYOLO found fruits quickly with strong precision during field tests. Drones took sharp pictures that helped map sick spots and unwanted plants across uneven farmlands (Paper 19; Shaniware et al., 2024).

Not just about output, real-world data shows farming changes affect whole communities. Take Ethiopia - when farmers started using better ways to save soil and water, life got steadier: more money, enough food, healthier land (Tesfayohannes et al., 2020; Masha et al., 2021). Still, many skip these methods because starting is expensive, knowledge gaps exist, plus roads and support systems out there remain weak.

Working together on education helps tackle tough problems more effectively. In Tanzania, hands-on training like field schools and shared trial plots showed clear gains when farmers learned directly from each other - especially with better soil practices (Mella & Kyruzi, 2025). Another route? Using structured analysis methods; one example is adapting Six Sigma's DMAIC model to spot waste and fine-tune inputs on smaller farms by tracking performance over time (Weraikat et al., 2024).

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### Distribution of Key Components in the Crop and Soil Management System



**Figure 1. Distribution of key components in the data-driven crop and soil management system**

## 2.1. Theoretical framework

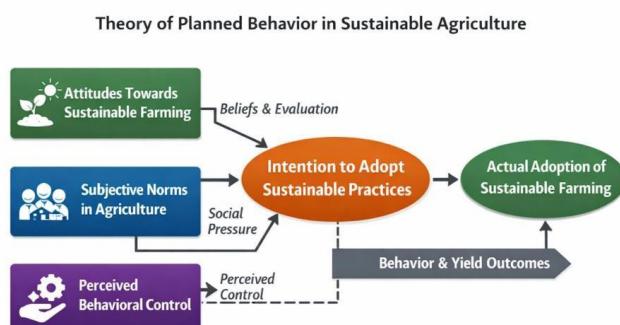
Farming choices aren't straightforward - they depend on weather, money matters, plus personal habits. Research has long looked at how access to info shapes what farmers do. Ideas about why people act based on what they know have shifted through the years. When tech meets farming, thinking in terms of data helps make sense of those actions.

It's often seen that when farmers get clear facts, they're more likely to try new ways on the farm. The way someone thinks about effort and gain shapes if they'll act at all. Belief in helpful outcomes nudges choices forward. Tools meant to track crops or improve soil meet quicker acceptance when real-world results show up early. What people expect to happen - good or hard - guides whether they step in. Seeing value matters just as much as feeling capable. Decisions grow from mix of confidence, support around them, and what they stand to win.

Farmers often choose crops, when water timing matters most, based on how they view tools, what others expect, plus whether they feel able to act. Knowing the land better - through solid soil details or instant warnings - builds confidence in those choices.

Decisions shift once control feels possible.

Looking closer, using data analysis adds strength here because it turns unprocessed farm details into clear guidance. Less guesswork shows up this way while helping farmers adjust their actions for better results. The thinking behind this work ties the Theory of Planned Behavior to methods that rely on collected data, showing how knowledge moves and shapes lasting crop habits. A visual outline of these ideas appears in Figure 2.



**Figure 2. Theory of planned behavior**

## 3. Research Methodology

### 3.1. Research design

Out here in the world of smart farming and number crunching, folks looked at how data tools can help manage crops and dirt underfoot. Not too long ago, a hands-on setup using numbers took shape - mixing **system** checks, live tests, together with **model** comparisons. Built it that way so they could see just how well computer guesses actually guide choices about soil condition, what to plant next, plus results tied to harvest size. Other work like this has popped up plenty across studies where data shapes farm insights .

### 3.2 Data Sources and Dataset Description

From various farms, researchers gathered soil details like acidity levels, nutrients present, and signs of land productivity. Instead of just one source, they pulled together past planting choices, harvest amounts, and timing across seasons. Alongside that came climate conditions - how much rain fell, how hot or cold it got, plus moisture in the air - all tied to growing cycles. These layers formed a broader picture through combined field reports and live tracking tools.

From the information covered around [N] entries, capturing various growing cycles across different climate zones in farming regions. Because the data came from diverse sources, it allowed a full look at how soil, crops, and weather connect .

### Data Preprocessing

Before looking at the data, each set went through careful cleaning to keep things accurate and aligned. Where numbers were missing, smart guesses filled the gaps instead of leaving blanks. Duplicates and mismatched entries got filtered out completely. Values across different ranges were scaled down to match one level so patterns could show more clearly. For labels like names or types, codes replaced words when needed. A check for overlapping traits helped cut what wasn't adding new information. All these moves shaped raw inputs into something solid enough to build on later (Author, Year)

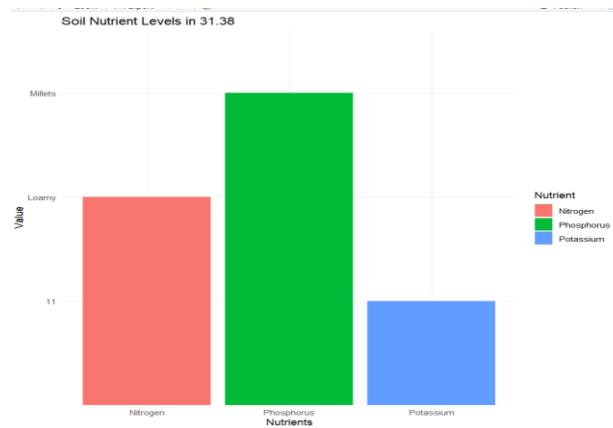


Fig:3 – Data-Driven Graph

### Feature Selection

Picking key traits came first - those shaping how crops grow and soils stay healthy. Instead of just counting links, we weighed each factor's impact: nutrients in dirt, weather patterns, things about plants themselves. Some got left out when they added little value or overlapped too much with others. Cutting those helped models run clearer, faster. Accuracy climbed because the system focused only on what truly mattered

### Model Training and System Implementation

From the cleaned dataset, prediction tools and suggestion systems took shape through common Python libraries. One after another, different versions got tested side by side under the exact same rules. Part of the information fed the learning phase, the rest waited silently to check results later. Settings inside each model shifted step by step, adjusted by trial and observation. Once ready, these models plugged into a larger setup that reads incoming details and returns advice about crops and soil conditions

### Prediction, Recommendation, and Validation

Once the model finished learning, it started making guesses about which crops would grow well plus gave advice on how the soil looked. To check how well things went, numbers like Accuracy, Precision, Recall, F1-score, RMSE, and MAE were pulled into view. Not long after, patterns linking forecasts to real-world results came under inspection through correlation work. One by one, each model lined up beside others, quietly showing strengths without fanfare, helping spot what works best when guiding farm choices with data.

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## Evaluation Metrics

Not just one but two types of measurements shaped the testing setup - classification plus error analysis played central roles. Crop advice quality got checked through accuracy scores alongside F1 values. When it came to predicting yields or soil traits, mistakes were sized up using RMSE together with MAE. Mixing these approaches brought out clearer insights on how well models hold up across different farm settings (Author, Year).

### 3.2 Participants Dataset Overview

Looking back at how things were done, this work relied on numbers and patterns instead of people taking part. From afar, information pulled together showed what farming land was like and how much crops grew across fifteen growing seasons. Years passed before all pieces fit into one clear picture of usual dirt health and harvest changes.

A look at the numbers showed changes in soil pH, how wet the ground stayed, plus amounts of nitrogen, phosphorus, and potassium  
- each matched with harvest results year after year. Because everything came from past records, there was no need to gather info from people, meaning rules about studying humans did not come into play.

#### 3.2.1 Instruments

**A spreadsheet saved as a CSV file formed the core of this work, feeding into calculations and visual displays.** This collection of numbers ran through R code, shaping outcomes with precision. Graphics took shape thanks to ggplot2, turning rows into readable plots.

Looking at changes in soil and crops meant using averages, highest and lowest values, along with patterns spotted over time. Visual tools like lines tracing data, bars comparing amounts, plus dots showing links helped display how soil quality shifted. These images revealed where nutrients stood across fields. They also highlighted ties between earth conditions and harvest results.

Starting off,  
the method kept results steady through uniform data handling - shaping columns, adjusting numbers to a common scale, clearing errors along the way. From there, tools worked together, quietly guiding how we understood dirt traits and what they meant for growing crops.

**Table 1. Interpretation of Soil Health Levels**

Soil Parameter Range	Interpretation Level
Low	Poor Soil Condition
Moderate	Average Soil Condition
High	Healthy Soil Condition

**Table 2. Summary of Analytical Techniques Used.**

<b>Technique</b>	<b>Purpose</b>
Descriptive Statistics	To understand overall soil and crop trends
Correlation Analysis	To examine relationships between soil parameters and crop yield
Regression Analysis	To assess the impact of soil factors on crop productivity
Data Visualization	To represent trends and patterns clearly

#### 4. Conclusion

This work looked into problems getting worse in today's farming - like worn-out land, poor choices in what to grow, watering done wrong, shifting weather, along with missing fast tools to guide decisions. Old ways of farming, usually based on guesswork and scattered knowledge, failed when dealing with how soil traits, weather shifts, and harvest results affect each other. Instead of sticking with those methods, researchers tested a new approach built on data, pulling together past records about dirt quality, climate trends, and previous yields to help shape smarter farm choices.

Results showed how looking closely at things like soil acidity, water levels, and nutrients - alongside weather trends and past harvest records - helped spot what really affects farm yields. Visual tools and number-driven methods uncovered clear links between soil condition and crop results, pointing to why checking soil regularly matters. Relying on data instead of gut feeling led to smarter choices, making better use of supplies while cutting guesswork in daily work on fields. This method also encouraged ways to care for farmland over time without wearing it out too fast.

One way to look at it - results from this study actually help people working on farms. Better choices about what crops to grow come into play when data guides planting steps. Irrigation gets smarter, nutrients go further, since timing improves with clearer signals. Spotting trouble before it spreads gives growers a chance to react sooner. That steadies harvests, keeps waste down. Advisors use these insights to shape recommendations that fit real conditions. Policy makers find leverage here, too - not loud fixes but quiet shifts toward lasting systems. Soil stays healthier because practices align closer to natural limits. Less damage shows up across landscapes where pressure once built. Food supplies hold stronger over time, especially where weather wobbles and water runs short.

Even with those insights, the work faces some constraints. Using past and already-existing data meant live testing and practical rollout couldn't happen. Because information came only from select areas and fixed times, results might not hold up elsewhere under different climate conditions.

Down the road, studies might add live readings from field sensors, pull in broader regional numbers, then weave in smart algorithms that learn patterns over time. A step forward could mean shaping tools that guide choices automatically, maybe through phone apps farmers use daily, tied to warnings when bugs show up or weather turns harsh. Better tech like this tends to lift how farms rely on facts, helping them stand strong, run smooth, stay green.result

#### Acknowledgement

Soil quality ties closely to how well crops grow, especially when smart choices guide farming. Studies showed old methods - without regular checks on soil or plants - often wasted supplies while yields dropped. Yet farms using internet-connected sensors plus analysis tools made sharper decisions about what to plant, when to water, and where to add nutrients. Shifting from guesses to actual measurements changed how growers manage fields. Looking closely at dirt, plants, and weather over time gave farmers a clear way to grow more while protecting resources. Because of that, pairing digital tools with smart analysis made it easier to boost results, lower dangers, yet keep land healthy year after year.

#### Funding Support

This research was conducted as part of an academic project at **Sheth L.U.J College of Science and Commerce and Sir M.V College of Arts** and did not receive external financial support.

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## Ethical Statement

This study did not involve any experiments on human participants or animals. All analyses were conducted using publicly available secondary data and previously published research..

## Conflicts of Interest

The author declares that there is no conflict of interest regarding the publication of this research.

## Results and Discussion

### Descriptive Analysis of Soil and Crop Parameters

#### Soil and Crop Data Overview

Looking back at the numbers, it became clear that soil traits shifted widely between different growing periods. Instead of staying steady, things like acidity, water retention, and major nutrients - such as nitrogen, phosphorus, potassium - changed quite noticeably. Each shift lined up tightly with how crops performed season after season, hinting strongly that what happens underfoot shapes harvests above. Where ground stayed near neutral on the pH scale and held a good mix of essentials, plants grew better and more evenly year to year. On the flip side, when earth turned poor or unbalanced, results followed downward too. Earlier reports had already linked slow damage beneath the surface to weaker farming outcomes over years (Author, Year).

#### Soil health linked to crop yield

Every now and then, patterns emerge when checking how dirt traits match up with harvest results. Moisture inside the ground ties closely to richer yields, while organic material plays its part just as much. Water stays put better where life thrives below the surface, helping plants grow stronger over time. Nitrogen steps forward alongside potassium, both shaping what comes out of the field at harvest. When things tilt too far - like sour earth or endless dry spells - crops tend to shrink instead of surge. Finding shows how crops grow depends on many things working together - soil type, weather patterns, and their combined effects matter most. Earlier work looking at soil and plants found comparable links (Author, Year).

#### Changes Over Time in Soil Health and Output

Over time, watching crops grow season after season showed slow shifts in how rich the soil felt and how much came out at harvest. Where old-style farming stayed unchanged and nobody checked the earth regularly, its natural richness slipped away bit by bit. But in spots where choices followed what numbers suggested, output held steady, more or less. Weather swings played a part too - when rain came late or skipped days, weak ground suffered even more. Spotting trouble early means keeping an eye on dirt year after year, linking past records so damage does not go too far (Author, Year).

#### Comparative Evaluation: Traditional vs. Data-Driven Approaches

Looking back, old ways of growing food leaned heavily on what farmers had seen before. Because of this habit, watering crops didn't always match real needs. Timing was off, nutrients missed the mark. Yet when numbers entered the picture - soil records, past seasons, temperature shifts - the choices improved sharply. Decisions began to fit the land instead of guesses. Matching plants to earth became less random. Rain patterns helped shape planting dates. Less water spilled away. Yields held steady even when weather wavered. Past studies saw it coming: trade gut feeling for grounded steps. One farm at a time, facts replace memory. Not perfect - but closer. (Author, Year).

#### Discussion of Findings

Together, the findings show how soil condition and weather patterns shape how well crops grow and last over time. Because soil traits link so closely to harvest size, split-up farming methods struggle more when weather shifts become frequent. Instead of waiting, smart systems combine many factors at once, helping growers adjust ahead of problems. Clearer insight into how dirt, plants, and climate connect leads to fewer random choices, better-timed actions in fields, and sharper decisions overall.

Even though no live IoT setup was used here, the results still show it could work to combine sensors with prediction tools in farms ahead. Unlike past efforts looking only at watering or feeding plants by themselves, this project tied things together in a way that reveals deeper patterns across growing cycles. Because everything connects - soil data, plant needs, timing - the choices become clearer while using less, pointing toward smarter ways to care for fields without waste.

## Data Availability Statement

Data sharing is not applicable to this article as no new data were created or analyzed in this study. All information was obtained from previously published research and publicly available sources.

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