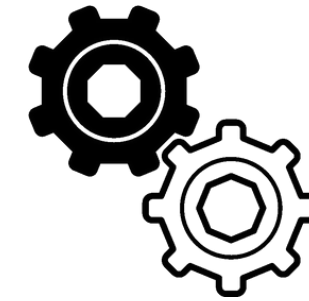


# SMART INDIA HACKATHON 2025



- **Problem Statement ID** – 25099
- **Problem Statement Title-** AI-powered monitoring of crop health, soil condition, and pest risks using multispectral/hyperspectral imaging and sensor data.
- **Theme-** Agriculture, FoodTech and Rural Development
- **PS Category-** Software
- **Team ID-** 102558
- **Team Name (Registered on portal) -** \_FarmAssist\_



We present an AI-powered agricultural platform that fuses hyperspectral imaging, IoT sensor data, and deep learning to provide real-time insights on crop health, soil quality, and pest risks. Built with Python, MATLAB, Next.js, and Firebase, it delivers spectral maps, soil summaries, and predictive alerts through an interactive dashboard. This unified system enables proactive crop management, enhancing productivity, reducing losses, and promoting sustainable farming practices.

KEY FEATURES

AI-Powered Analysis

Uses CNN and LSTM models to interpret fused hyperspectral and sensor data with greater accuracy.

01



Interactive Dashboard

Offers health maps, soil summaries, and charts that are easy to navigate for farmers and agronomists.

02



03

Zone-Specific Guidance

Delivers customized recommendations for each section of a field, improving efficiency and productivity.



04

Predictive Trend Insights

Tracks crop growth, soil shifts, and risk zones with predictive alerts for timely interventions.

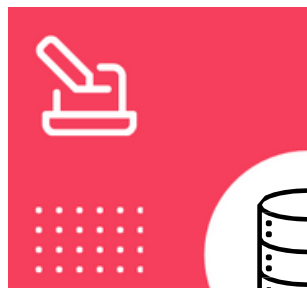


INNOVATION AND UNIQUENESS

Fusion of Data Sources

Combines hyperspectral imaging with IoT sensor data for richer and more reliable insights.

01



02

Farmer-First Design

Simplifies complex analytics into actionable insights without overwhelming users with technical details.



03

Continuous Learning Models

Adapts over time to local conditions, improving prediction accuracy with every season.



04

Proactive Crop Management

Transforms farming from reactive problem-solving to proactive planning for higher yields and sustainability.



HOW IT ADDRESSES THE PROBLEM

01

Unified Field Monitoring

Integrates satellite imagery and ground sensors to give farmers a complete, real-time view of their fields.



02

Early Stress Detection

Identifies signs of crop stress, pests, and diseases before they become visible, enabling preventive actions.



Reduced Manual Effort

Automates data gathering and analysis, saving time and resources compared to traditional scouting.

04

Data-Driven Farming

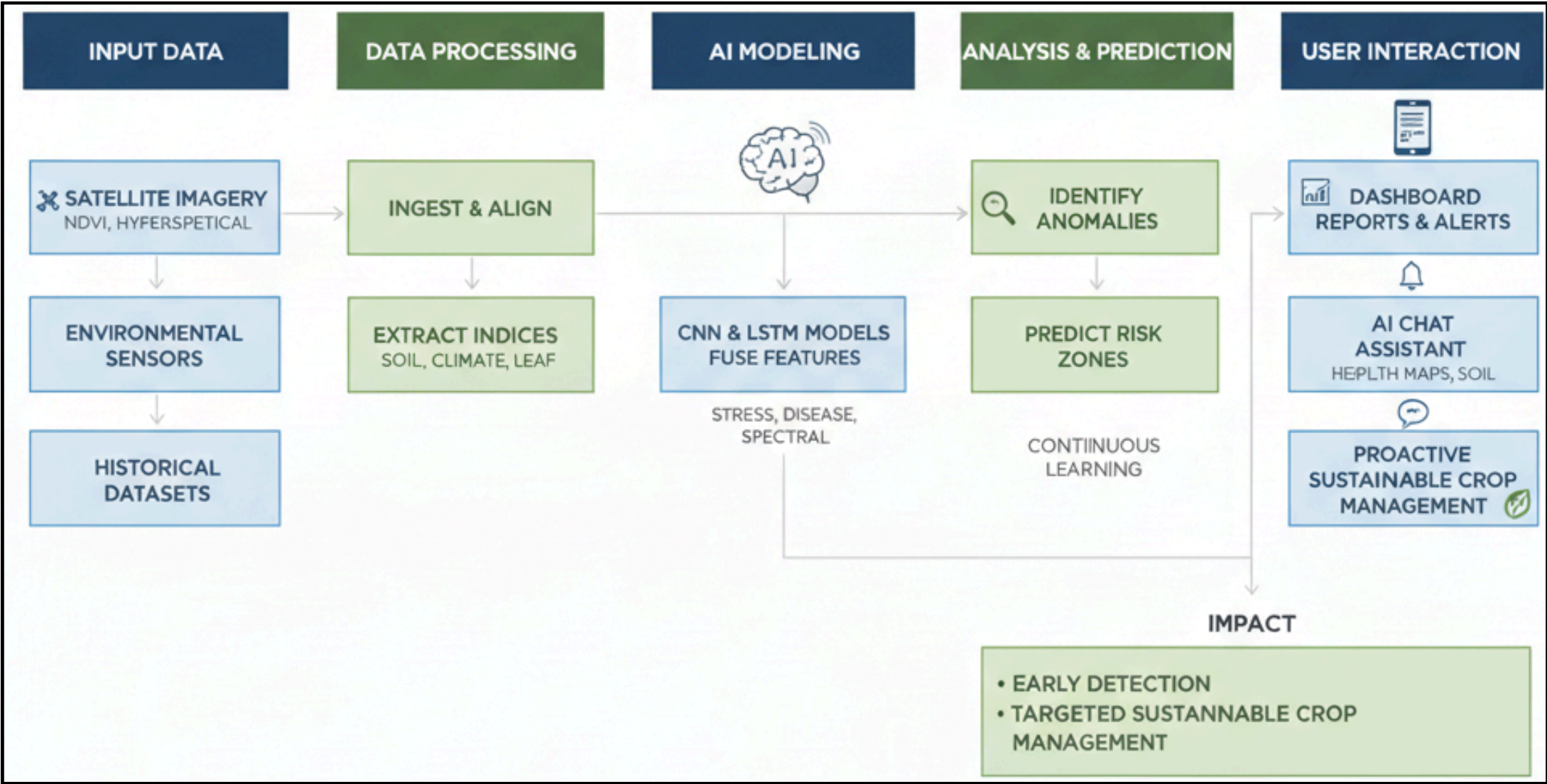
Provides precise recommendations that help optimize irrigation, fertilization, and pest control decisions.



03



## Digital Crop Analysis Workflow



## Crop Management Workflow

Agricultural data is vast and complex. Our system streamlines the process, making crop insights faster, reliable, and farmer-friendly. Its advantages are:

- Automated Preprocessing:** Cleans and structures raw data for accurate results.
- Scalable Handling:** Manages large datasets from sensors and satellites with ease.
- Smart Predictions:** Detects crop stress, pests, and yield risks early.
- Interactive Visualization:** Provides clear maps, alerts, and soil condition summaries.
- Optimized Workflow:** Delivers faster, reliable insights for better farm decisions.

## Precision farming workflow



### Programming Languages:

Python and MATLAB → For ML model development & preprocessing  
TypeScript → For frontend development

### Frameworks & Libraries:

Next.js & React → Scalable frontend framework & UI components  
Tailwind CSS & Radix UI → Styling and accessible UI primitives  
Recharts → Data visualization & analytics  
MATLAB (Hyperspectral Imaging, Image Processing, Deep Learning Toolbox) → Spectral data analysis & model training  
PyTorch → ML/DL model training & deployment  
Firebase - For data storage



## CHALLENGES AND RISKS



### Alignment & Synchronization

GPS errors and time mismatch between satellite images and ground sensor readings.



### Environmental Factors

Clouds, haze, and poor image quality reduce spectral data reliability.



### Crop & Soil Variability

Different crop types, soil conditions, and sensor depths affect model accuracy.



### Computational & Expertise Limits

Deep learning models need high compute power and domain knowledge.

## STRATEGIES TO OVERCOME CHALLENGES



### Data Harmonization

Aggregate small patches and allow tolerance ( $\pm 2$  days,  $\leq 30$  m) for accurate alignment.



### Quality Control

Apply cloud masks, use quality flags, and collect multi-temporal images to minimize gaps.



### Model Customization

Train crop/soil-specific models or generalize with larger, diverse datasets.



### Progressive Scaling

Begin with simpler models (RF, SVM) and gradually scale to CNN/LSTM with more resources.

## FEASIBILITY

### Technical Feasibility



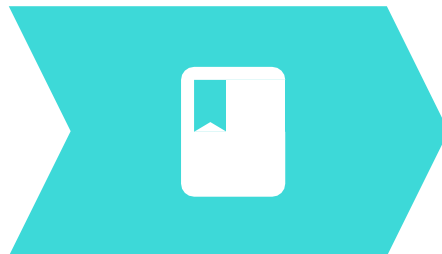
The solution leverages existing tools (MATLAB/HSI Library, Sentinel-2, sensor data) and proven AI models (SVM, CNN, LSTM), making implementation technically achievable.

### Operational Feasibility



Field sensors, remote sensing data, and farmer-friendly dashboards ensure smooth adoption, with minimal disruption to existing farming workflows.

### Economic Feasibility



Open-source datasets, scalable cloud resources, and gradual AI model upgrades reduce costs, making the solution cost-effective and sustainable for long-term use.



## POTENTIAL IMPACT ON THE TARGET AUDIENCE




### Improved Decision-Making

Farmers get clear, data-backed insights to act at the right time.




### Risk Reduction

Early detection of crop stress and pests lowers chances of major yield losses.



### Resource Optimization

Efficient use of water, fertilizer, and pesticides saves costs and land health.



### Higher Yield & Income

Stronger crops and timely interventions boost productivity and profits.

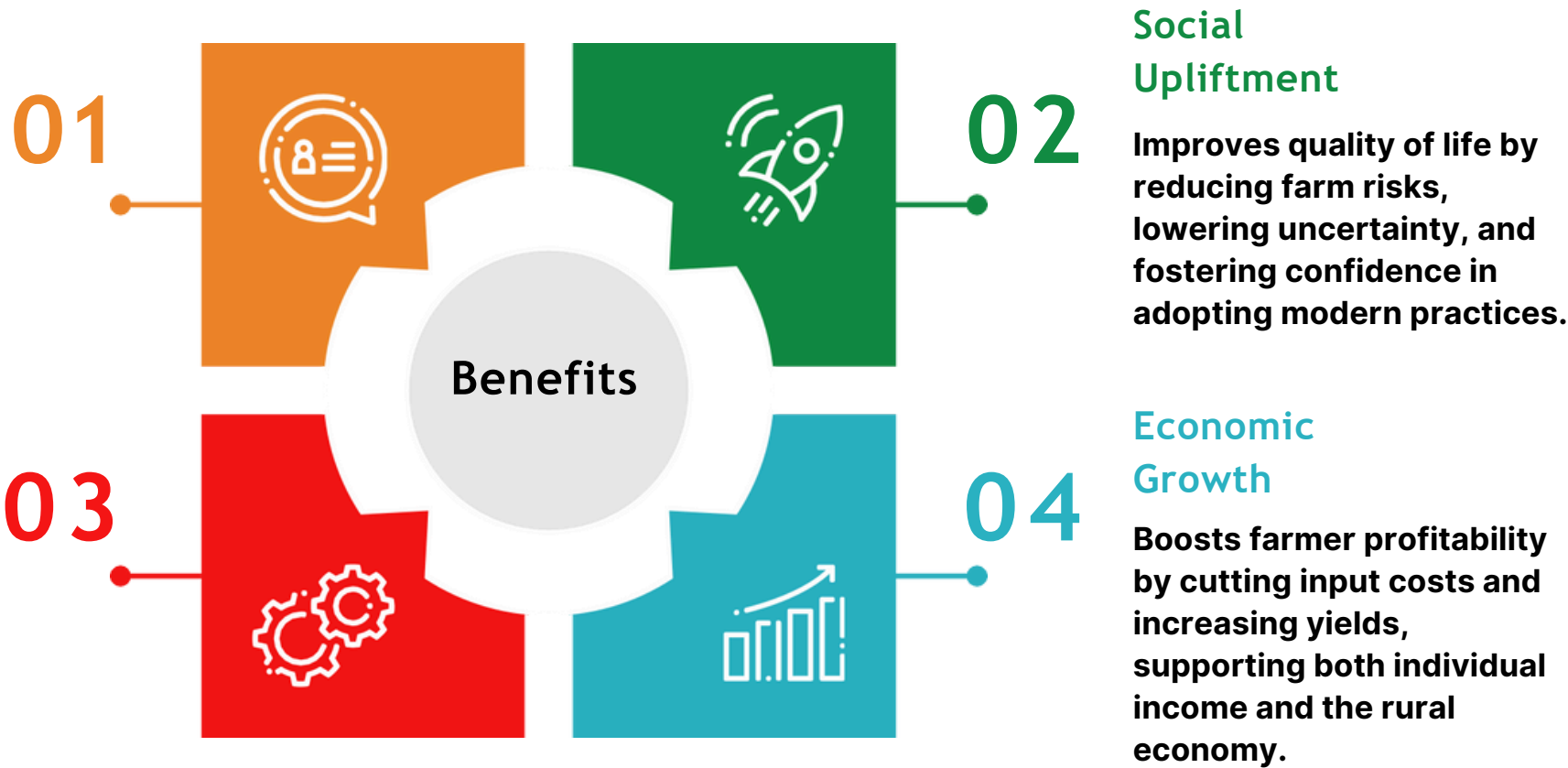
## BENEFITS OF THE SOLUTION

### Innovation in Agriculture

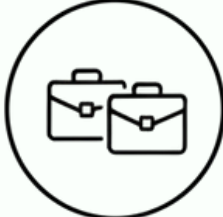
Integrates AI, remote sensing, and IoT to modernize farming practices, paving the way for scalable precision agriculture and long-term resilience.

### Environmental Protection

Reduces water wastage, lowers fertilizer runoff, and preserves soil health, contributing to long-term ecological balance.




## REVENUE MODEL




### B2B / Enterprise Licensing

Platform access for agri-input companies, crop insurers, & govt. bodies




### Insurance & Risk Assessment

Predictive risk insights for insurers (pricing & claims)



### Subscription Plans (SAAS)

Monthly/annual fees for dashboards, reports, & alerts



### Freemium → Premium Upsell

Basic features free; premium AI predictions & advice

# **RESEARCH AND REFERENCES**

## **Live Deployment:**

<https://crop-health-dashboard-ten.vercel.app/>

## **Demo Video:**

<https://youtu.be/dOyprAfEQ2k>

## **Source Code:**

[https://github.com/Akshat-kacodia/SIH\\_2025\\_Submission](https://github.com/Akshat-kacodia/SIH_2025_Submission)

## **Research Basis:**

- Hyperspectral Imaging for Crop Stress Detection → Used to identify early signs of disease & nutrient deficiency.  
(Mahlein et al., 2018 – Hyperspectral Imaging for Precision Agriculture)
- Soil Moisture & Vegetation Health Link → NDVI & spectral indices influenced by soil moisture levels.  
(Gao et al., 2020 – Monitoring Soil Moisture using Remote Sensing)
- AI for Early Detection & Prediction → CNN, LSTM, SVM effective in classifying stress & spectral patterns.  
(Kamilaris & Prenafeta-Boldú, 2018 – Deep Learning in Agriculture)

## **Citations:**

- Mahlein, A.K. et al. (2018). Hyperspectral Imaging for Precision Agriculture.
- Gao et al. (2020). Monitoring Soil Moisture with Remote Sensing.
- Kamilaris & Prenafeta-Boldú (2018). Deep Learning in Agriculture: A Survey.
- Indian Pines Dataset – Purdue University.
- Victoria Soil Sensor Open Data Portal.