

Farmer Assistant Application

A project report submitted in fulfillment of the
requirements for the degree of

Bachelor of Computer Engineering (Sem -VIII)

by

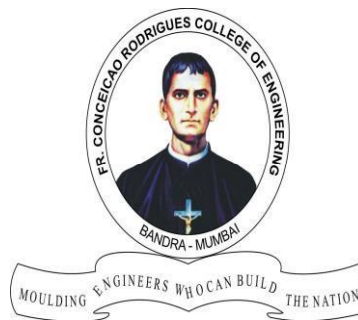
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(2022-2023)

This work is dedicated to my family.

I am very thankful for their motivation and support.

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CERTIFICATE

This is to certify that the project entitled "**Farmer Assistant App**" is a bonafide work of **Pratik Harde (8871), Ibin Babu (8872), Ananya Sharma (8908)** submitted to the University of Mumbai in partial fulfillment of the requirement for the award of the degree of **Bachelor of Engineering in Computer Engineering (Sem VII)**.

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Date: 29 March 2023

Place:

Declaration

We declare that this written submission represents our ideas in our own words and where others' ideas or words have been included, we have adequately cited and referenced the original sources. We also declare that we have adhered to all principles of academic honesty and integrity and have not misrepresented or fabricated or falsified any idea/data/fact/source in my submission. We understand that any violation of the above will be cause for disciplinary action by the Institute and can also evoke penal action from the sources which have thus not been properly cited or from whom proper permission has not been taken when needed.

Pratik Harde (Roll No. 8871) (**sign**) _____

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Date: 29 March, 2023

Abstract

In India, agriculture is a significant source of both income and employment. The most frequent issue that Indian farmers encounter is that they choose the incorrect crop for their soil, use the incorrect fertiliser, and are unable to identify plant diseases that are caused by their plants. As a result, they will see a major decline in productivity. Farmers' problems will be resolved via the Farmer Assistant App. In order to suggest the optimum crop to farmers, as well as fertiliser suggestions based on site-specific characteristics, precision agriculture uses research data on soil qualities, soil types, and crop production statistics. By doing this, crop selection errors become less frequent and productivity rises. The proposed recommendation and prediction system in this study uses ML models as learners to highly accurately and effectively select a crop for the site-specific factors. The mechanism for recommending fertiliser is also entirely based on Python logic. Here, we compare the data (the crop's ideal nutrients) with the user-entered information. The most variable nutrient is then classified as HIGH or LOW, and recommendations are then retrieved in accordance with those results.

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Glossary

SVM Support Vector Machines

RF Random Forest

DFD Data Flow Diagrams

ML Machine Learning

KNN K-Nearest Neighbor

Chapter 1

Introduction

A farmer's decision regarding which crop to cultivate is typically influenced by his knowledge as well as unimportant variables like the desire to make quick money, ignorance of market demand, exaggeration of a soil's ability to support a particular crop, and so forth. The farmer's family's financial situation could be severely strained if he makes the wrong choice. Maybe this is one of the numerous factors contributing to the innumerable farmer suicide cases that the media reports on every day. Such a wrong judgement would have detrimental effects on not only the farmer's family but the entire economy of an area in a country like India, where agriculture and associated sectors contribute to about 20.4% of its Gross Value Added (GVA). Because of this, we have determined that a farmer's decision on which crop to cultivate during a specific season is a very serious one. The urgent requirement is to create a system that might offer Indian farmers predictive information so they could choose which crop to produce with knowledge.

In light of this, we suggest a system, an intelligent system, that would evaluate soil characteristics (pH value, N, P, K, soil type, and nutrients concentration) as well as environmental factors (rainfall, temperature, geographic location in terms of state), before advising the user on the crop that would grow best. Additionally, a fertiliser recommendation based on the ideal nutrients of the produced crops is also made.

1.1 Motivation

As about 70% of the population depends on agriculture, it serves as the economic foundation for developing nations like India. India's economy and employment are heavily reliant on agriculture

Indian farmers frequently struggle with choosing the right crop to support their soil's needs and which fertiliser to utilise for their particular crop. This problem of the farmers has been addressed through precision agriculture

1.2 Project Scope

1. Crop Advice System: Develop a system that provides advice to farmers on the best crops to grow in the area based on the estimated crop sustainability. The system should be designed to prevent farmers from suffering any losses by providing them with information on the most suitable crops for their specific area.
2. Fertilizer Recommendation System: Develop a system that recommends fertilizers for specific crops based on their chemical properties and the identified soil and climatic conditions. The system should be designed to improve crop yields while minimizing the use of harmful chemicals.
3. Disease Detection System: Develop a system that uses machine learning algorithms to detect common crop diseases based on symptoms and other relevant data. The system should provide recommendations for disease control and prevention.

1.3 Objectives

- To develop a reliable model that can accurately estimate crop sustainability in a given state under specific climatic and soil conditions
- Give advice on the best crops to grow in the area so that the farmer doesn't suffer any losses.
- Suggest fertilizers for crops with supported chemical properties.

Chapter 2

Literature Review

Low-cost IOT + ML design for smart farming with multiple application paper authors Fahad Kamraan Syed, Agniswar Paul, Ajay Kumar, Jaideep Cherukuri in paper [1] proposed system for water management systems and improve current irrigation methods. An IoT and MLbased farming system always keeps farmers aware of the upcoming weather possibilities and gives them the best suggestions about irrigation methods and crops thereby helping in better yield.

In paper [2] author's proposed a smart system that can assist farmers in crop management by considering sensed parameters (temperature, humidity) and other parameters (soil type, location of farm, rainfall) that predicts the most suitable crop to grow in that environment.

Reference Paper [3] determines real time sampling of soil properties using MODIFIED SUPPORT VECTOR REGRESSION, a popular machine learning algorithm and four modules. The Modules include Sensor interfaced to IoT device, Agri cloud, Analyzing the real time sensor data and Agri user interface (AUI). The first module is portable IoT device (NodeMCU) with soil moisture sensor and pH sensor, environmental sensors. Agri cloud module consists of storage. Analyzing the real time data module is processing of types of crops and small plants suggested using modified support vector machine algorithm. Agri-user interface is a basic web interface. Thus, with the help of soil properties farmer will be able to get types of crops and small plants is grown in farmland with help of Modified support vector machine algorithm.

In paper [4] author's proposed new technologies include the use of Internet of Things (IOT) and Machine Learning. The real time data from the field area can be collected using IOT system. The collected data from the field area is fed to the trained model. The trained model then makes the predictions using the data. The result produced by the model greatly helps in sowing the suitable crops in the particular field area. GHRIET, Department of Computer Engineering 2020-21 10

In Reference paper [5] determines a model is proposed for predicting the soil type and suggest a suitable crop that can be cultivated in that soil. The model has been tested using various machine learning algorithms such as KNN, SVM and logistic regression. The accuracy of the present model is maximum than the existing models.

Aruul Mozhi Varman S proposed an IOT and deep learning based smart agriculture systems. This system monitors and collects the soil parameters from the field with the help of a wireless sensor

network. The collected data is then uploaded in the cloud. Finally, the systems suggest best irrigation practices to the farmers by predicting the crop to be sown for next crop rotation. This information will be sent as an SMS to the farmers. The parameters include soil temperature, atmospheric temperature, and humidity [6]. This system suggests further improving the effectiveness by predicting the suitable time for applying pesticides, fertilizer and manures.

In paper [7] proposed a system would assist the farmers in making an informed decision about which crop to grow depending on a variety of environmental and geographical factors. The ML and IoT based suggestions will significantly educate the farmer and help them minimize costs and make strategic decisions by replacing intuition and passed-down knowledge with far more reliable data-driven ML models. This allows for a scalable, reliable solution to an important problem affecting hundreds of millions of people

In paper [8], proposes a machine learning-based approach for predicting the onset of various diseases based on various features such as NPK value, Temperature, etc. The authors utilize six different machine learning algorithms, including Decision Tree, Random Forest, K-Nearest Neighbor, Naive Bayes, Support Vector Machine, and Logistic Regression. The dataset used for this study includes 768 classes from the Kaggle dataset, and the results show that the Random Forest algorithm achieved the highest accuracy of 77.60%.

In paper [9], proposes a machine learning-based approach for detecting plant diseases using leaf images. The authors utilize three different machine learning algorithms, including Convolutional Neural Network (CNN), Support Vector Machine (SVM), and K-Nearest Neighbor (KNN), to classify healthy and diseased plant leaves based on the image features extracted from them. The results show that the CNN algorithm achieved the highest accuracy of 98.45% in classifying the plant leaves into their respective disease categories.

In paper [10], proposes a data mining-based approach for recommending suitable crops and fertilizers for a particular agricultural region. The authors utilize various data mining techniques, including association rule mining, decision tree, and k-means clustering, to analyze and process the agricultural data. The study also recommends the appropriate fertilizers for the recommended crops based on the association rules mined from the dataset. The study concludes that data mining techniques can effectively assist farmers in making informed decisions about crop selection and fertilizer application, ultimately leading to higher yields and economic benefits.

In paper [11], proposes a machine learning-based approach for recommending suitable crops for a particular agricultural region. The authors utilize various machine learning algorithms, including K-Nearest Neighbor (KNN), Decision Tree, and Random Forest, to analyze and process the agricultural data. The study also considers various factors such as pH level, rainfall, temperature, and humidity in recommending suitable crops for a particular region. The study concludes that machine learning algorithms can effectively assist farmers in making informed decisions about crop selection, ultimately leading to higher yields and economic benefits.

2.12 Literature Survey Table

No.	Reference	Dataset	Techniques	Results
1.	An IoT Based Smart Farming System Using Machine Learning (researchgate.net)	The data have been collected from 30 US and Canadian Cities, as well as 6 Palestine cities. The dataset contains ~5 years of high temporal resolution (hourly measurements) data of various weather attributes, such as temperature, humidity	Data collection phase using sensors deploying in an agricultural field. Data cleaning and storage phase and predicting processing using some AI methods.	Showned that AI techniques play a pivotal role in agriculture of precision by using machine learning and open sources technologies.
2.	Smart Management of Crop Cultivation using IOT and Machine Learning IRJET Journal - Academia.edu	-	uses random forest algorithm to predict appropriate crop based on current NPK value of soil. Machine Learning Algorithm(KNN) is used to calculate the crop which is best to grow in the particular field based on the values received at real time.	Using the real time values obtained from the field and running the algorithm on them, the most suitable crop for a particular land at a given time is predicted

3.	High Resolution Mapping of Soil Properties Using Remote Sensing Variables in South-Western Burkina Faso: A Comparison of Machine Learning and Multiple Linear Regression Models (researchgate.net)	A total of 1104 soil samples (1002 in subwatershed and 102 outside)		Internal validation was conducted by cross validation while the predictions were validated against an independent set of soil samples considering the modelling area and an extrapolation area.
4.	Internet of things (IoT) applications to fight against COVID-19 pandemic - PMC (nih.gov)	No Dataset information is provided	IoT is used to capture health data from various locations of the infected patient and manage all the data using the virtual management system contact tracing, cluster identification and compliance of quarantine.	By using a statistical-based method, IoT gets helpful to predict an upcoming situation of this disease.
5.	Soil Classification and Crop Suggestion using Machine Learning Techniques by IJRASET - Issuu	They have used a large dataset extracted from the Australian Department of Agriculture and Food(AGRIC) to conduct the research	The model has been tested using various machine learning algorithms such as KNN, SVM and logistic regression.	The accuracy of the present model is maximum than the existing models.
6.	Smart Agriculture Using WSN and IoT: Environment & Agriculture Book Chapter IGI Global (igi-global.com)	-	This system monitors and collects the soil parameters from the field with the help of a wireless sensor network. The collected data is then uploaded in the cloud. Finally, the systems suggest best irrigation	This system suggests further improving the effectiveness by predicting the suitable time for applying pesticides, fertilizer, and manures.

			practices to the farmers by predicting the crop to be sown for next crop rotation.	
7.	Intelligent Agriculture System To Assist Farmers In Smart Decision (quickcompany.in)	A total of 9 types of soil present in dataset.	The ML and IoT based suggestions will significantly educate the farmer and help them minimize costs and make strategic decisions by replacing intuition and passed-down knowledge with far more reliable data-driven ML models.	provide intelligent agriculture system to assist farmers in smart decision making using IoT data analytics and machine learning.
8	Disease Prediction From Various feature Using Machine Learning by Rinkal Keniya, Aman Khakharia, Vruddhi Shah, Vrushabh Gada, Ruchi Manjalkar, Tirth Thaker, Mahesh Warang, Ninad Mehendale :: SSRN	The dataset used had more than 230 diseases	The weighted KNN algorithm gave the best results as compared to the other algorithms. The accuracy of the weighted KNN algorithm for the prediction was 93.5 %.	Different machine learning models were used to examine the prediction of disease for available input dataset. We used 11 different ML models for the prediction. Out of the 11 models we managed to get 50 % or above accuracy for 6 models.
9	(PDF) Plant Disease Detection Using Machine Learning (researchgate.net)	No Dataset information is provided	Compared to other machine learning techniques like SVM, Gaussian Naïve bayes, logistic regression, linear discriminant analysis, Random	Random Forest gives highest accuracy of 70% and

			forests gave more accuracy with less number of image data set	
10	(PDF) Crop Suitability and Fertilizers Recommendation Using Data Mining Techniques (researchgate.net)	No Dataset information is provided	For recommending fertilizer to the farmer, K-means clustering algorithm is used. Crop name and soil contents Nitrogen (N), Phosphorous (P) and Potassium (K) are given as an input to the given algorithm.	Maximum accuracy of random forest is 71%
11	https://ijariie.com/AdminUPI/crop_recommendation_System_using_Mchine_Learning_ihariee174.pdf	"The dataset contains various parameters like Nitrogen (N), Phosphorous (P), Potassium (K), PH value of soil, Humidity, Temperature, and Rainfall	Various algorithms are used like KNN algorithm, random forest, support vector machine, XGBoost	Maximum accuracy is of Naive bayes 94.72% and minimum accuracy is of AdaBoost classifier with 12%.

Chapter 3

Problem Statement

The overall problem statement is to develop a comprehensive agricultural solution that encompasses crop recommendation, fertilizer recommendation, and disease prediction. The solution should provide accurate and personalized recommendations to farmers based on various environmental variables such as soil type, weather conditions, and historical data. The aim is to help farmers optimize their crop yields and minimize losses due to disease outbreaks or inadequate fertilizer application. The solution should be scalable, cost-effective, and accessible to farmers across different regions and socio-economic backgrounds.

3.1 Drawbacks of Current Solutions

The fact that each author of each publication concentrated on one factor (weather or soil) for predicting the appropriateness of crop development was one weakness we found in all of these important published papers. But in our opinion, the best and most accurate prediction should be made by concurrently taking into account both of these aspects. This happens frequently because even if a particular soil type may be ideal for supporting a certain crop variety, the yield will suffer if the local climate doesn't appear to be favourable for that crop kind.

3.2 Solution to the Problem

By proposing an effective crop recommendation system that accounts for all relevant factors, such as temperature, rainfall, location, and soil quality, we hope to eliminate the above shortcomings. Giving farmers crop suggestions is the main task that this system is primarily concerned with carrying out. In order to give the user a simple and trustworthy knowledge of the insight to determine and plant the crops, we also supply the fertilisers to be used for crops grown in various states.

Chapter 4

Project Design

4.1 Overview

Fig 4.1 shows the architecture of whole system and how the prediction happens from uploading the image to final result.

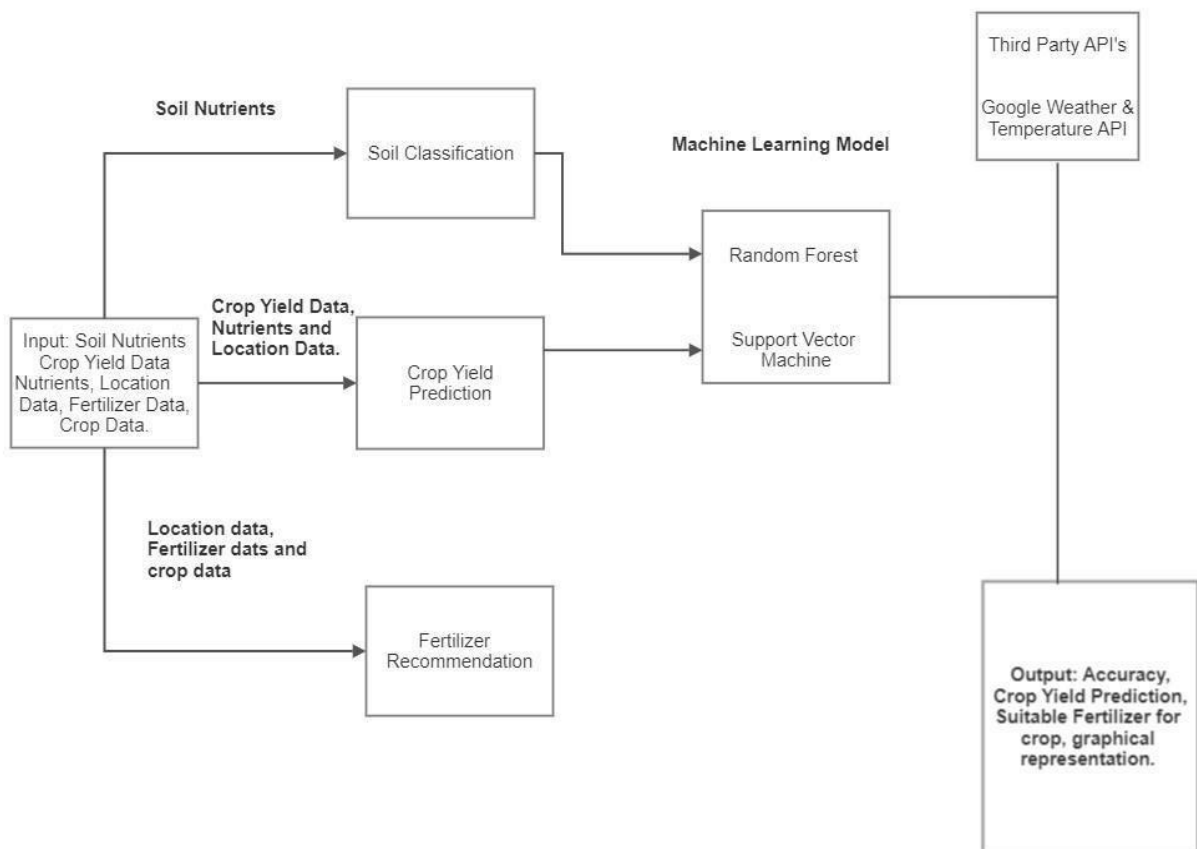


Fig 4.1 Block Diagram for Crop and Fertilizer Recommendation

Crop recommendation, fertiliser recommendation, and disease prediction are the three goals of our method. We are utilising an API key to retrieve the data for the fertiliser recommendation. The system will provide recommendations and predictions based on the nutrients dataset.

4.2 Architecture

Fig 4.2 gives ideas about architecture of system. How data is collected and how models are predicting result based on input data

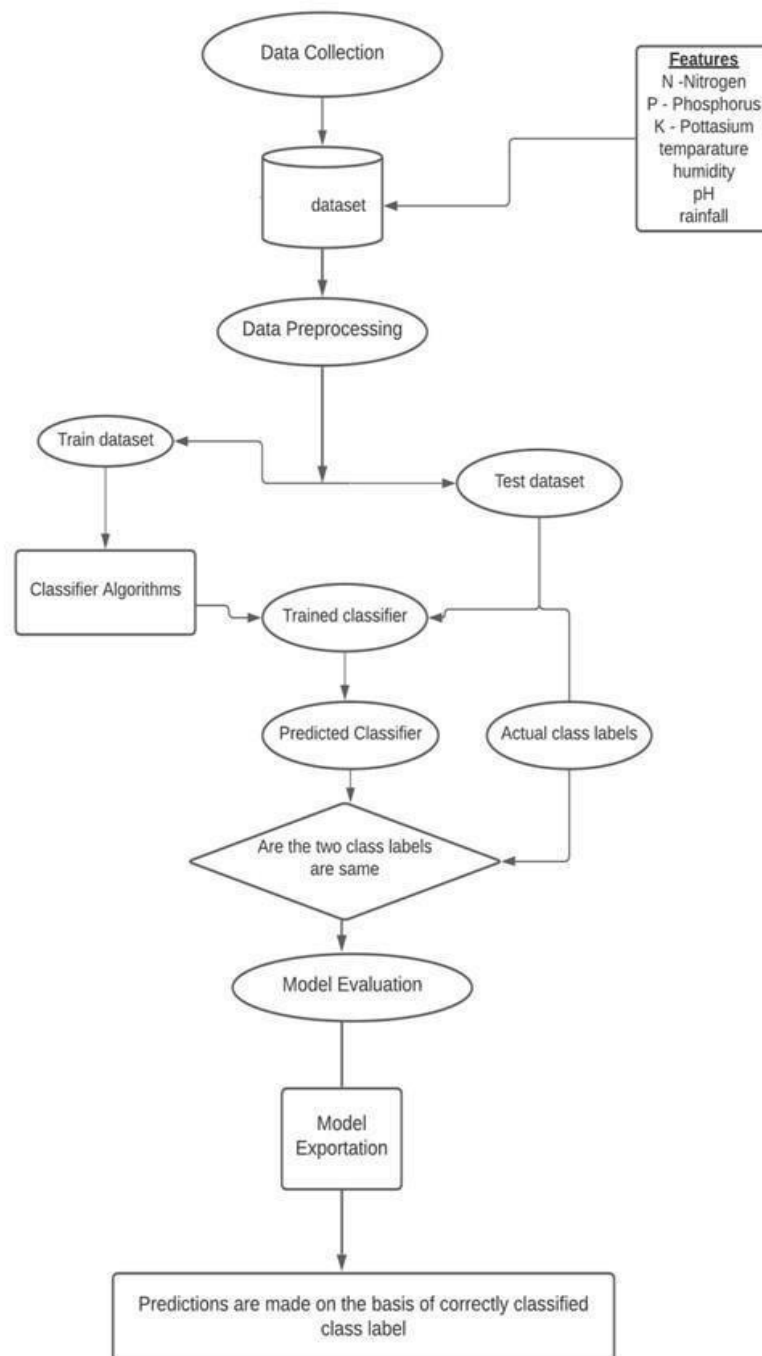


Figure 4.2 : Architecture

4.3 Module Description

Module Description for Crop Recommendation:

The crop recommendation module aims to recommend the most suitable crop to cultivate based on various environmental factors like soil type, climate, and water availability. The module utilizes data analytics and machine learning techniques to analyze data and provide accurate recommendations. The module considers various factors like the current season, crop rotation, and previous crop yield to provide optimal crop recommendations. The module provides farmers with a list of suitable crop options, along with information on their yield potential and growth requirements.

Module Description for Fertilizer Recommendation:

The fertilizer recommendation module aims to suggest the appropriate type and amount of fertilizer to apply to a crop based on various environmental factors like soil type, crop type, and weather conditions. The module utilizes machine learning algorithms to analyze soil samples and recommend the right type of fertilizer. The module also considers the crop's growth stage, previous fertilizer application, and expected yield to provide accurate recommendations. The module provides farmers with detailed instructions on the optimal fertilizer application method and schedule.

Module Description for Disease Prediction:

The disease prediction module aims to forecast the likelihood of crop diseases in a particular region or field. The module utilizes machine learning algorithms to analyze various environmental factors like temperature, humidity, soil moisture, and other indicators to predict the likelihood of disease occurrence. The module also provides farmers with recommendations on preventive measures to reduce the likelihood of disease outbreaks. The module also recommends suitable pesticides and treatments to manage the disease effectively. The module provides farmers with early warning alerts to take timely preventive measures and minimize crop losses due to disease outbreaks.

4.4 Algorithms

4.4.1 Decision Tree

A nonparametric supervised learning technique for classification and regression is called a decision tree (DT). The objective is to learn straightforward decision rules derived from the data features in order to build a model that predicts the value of a target variable. A fractional approximation of a tree can be thought of. For instance, in the example below, using a series of if-then-else decision rules, decision trees learn from data to approximate a sine curve.

4.4.2 Support Vector Machine

In an N-dimensional space (N being the number of features), the support vector machine algorithm seeks to locate a hyperplane that clearly categorises the data points.

There are a variety of different hyperplanes that might be used to split the two classes of data points.

Finding a plane with the greatest margin—that is, the greatest separation between data points from both classes—is our goal. Maximizing the margin distance adds some support, increasing the confidence with which future data points can be categorised.

4.4.3 Logistic Regression

When classifying observations into a collection of discrete classes, the classification algorithm logistic regression is utilised. Email spam or not spam, online fraud or not fraud, and malignant or benign tumours are a few instances of classification issues.

To convert its output into a probability value, logistic regression uses the logistic sigmoid function.

What are the types of logistic regression

1. Binary (eg. Tumor Malignant or Benign)
2. Multi-linear functions fails Class (eg. Cats, dogs or Sheep's)

The machine learning technique known as logistic regression, which is based on the probability notion and used to solve classification problems, is used to analyse data in a predicted manner.

4.4.4 Random Forest

Supervised machine learning algorithms like random forest are frequently employed in classification and regression issues. On various samples, it constructs a vote for regression.

The Random Forest Algorithm's ability to handle data sets with both continuous variables, as in regression, and categorical variables, as in classification, is one of its most crucial qualities.

In terms of classification issues, it delivers superior outcomes.

To precisely address the issue of high-variance in Decision Trees, Random Forests were created. You're not just training one Decision Tree, as the name implies—you're training an entire forest! A forest of Bagged Decision Trees in this instance.

Random Forests algorithm follows these steps:

- 1 Starting with the original dataset, generate N bagged samples of size n , where n is the number of samples to be created.
- 2 With each of the N bagged datasets as input, train a decision tree. Don't, however, look at every feature in the dataset while doing a node split. Choose M features at random from the entire training set of features. Then, using impurity metrics like Gini Impurity or Entropy, choose the optimal split.
- 3 Combine the outcomes of each decision tree into a single output.
- 4 If you're working on a regression problem, average the results for each observation, as produced by each tree.
- 5 If you're working on a classification problem, make a majority decision for each observation across all trees.

4.5 Dataset

Datasets include:-

1. Crop Recommendation dataset

2. Fertilizer dataset

3. Plant Disease dataset

4. State -District Wise Crop dataset A brief description of the datasets:

5. **Crop Recommendation dataset** : In kg per hectare, this information shows the yield for 18 important crops farmed in every state. When the yield is 0, it means that the crop is not grown in that stage. containing a total of 2201 items in the dataset. It includes 8 parameters i.e N, P, K, Temperature, humidity, pH, rainfall and label.

1	N	P	K	temperatu	humidity	ph	rainfall	label
2	90	42	43	20.87974	82.00274	6.502985	202.9355	rice
3	85	58	41	21.77046	80.31964	7.038096	226.6555	rice
4	60	55	44	23.00446	82.32076	7.840207	263.9642	rice
5	74	35	40	26.4911	80.15836	6.980401	242.864	rice
6	78	42	42	20.13017	81.60487	7.628473	262.7173	rice
7	69	37	42	23.05805	83.37012	7.073454	251.055	rice
8	69	55	38	22.70884	82.63941	5.700806	271.3249	rice
9	94	53	40	20.27774	82.89409	5.718627	241.9742	rice
10	89	54	38	24.51588	83.53522	6.685346	230.4462	rice
11	68	58	38	23.22397	83.03323	6.336254	221.2092	rice
12	91	53	40	26.52724	81.41754	5.386168	264.6149	rice
13	90	46	42	23.97898	81.45062	7.502834	250.0832	rice
14	78	58	44	26.8008	80.88685	5.108682	284.4365	rice
15	93	56	36	24.01498	82.05687	6.984354	185.2773	rice
16	94	50	37	25.66585	80.66385	6.94802	209.587	rice
17	60	48	39	24.28209	80.30026	7.042299	231.0863	rice
18	85	38	41	21.58712	82.78837	6.249051	276.6552	rice
19	91	35	39	23.79392	80.41818	6.97086	206.2612	rice
20	77	38	36	21.86525	80.1923	5.953933	224.555	rice

Fig 4.3 Crop recommendation dataset

- **Fertilizer dataset** : The columns in this dataset have the following attributes: State, Nitrogen, Phosphorous, Potassium, Average Ph Level, and Soil Moisture Content. There are 25 total records in the collection.

	Crop	N	P	K	pH	soil_moisture
0	rice	80	40	40	5.5	30
3	maize	80	40	20	5.5	50
5	chickpea	40	60	80	5.5	60
12	kidneybean	20	60	20	5.5	45
13	pigeonpea	20	60	20	5.5	45
14	mothbean	20	40	20	5.5	30
15	mungbean	20	40	20	5.5	80
18	blackgram	40	60	20	5	60
24	lentil	20	60	20	5.5	90
60	pomegranate	20	10	40	5.5	30
61	banana	100	75	50	6.5	40
62	mango	20	20	30	5	15
63	grapes	20	125	200	4	60
66	watermelon	100	10	50	5.5	70
67	muskmelon	100	10	50	5.5	30
69	apple	20	125	200	6.5	50
74	orange	20	10	10	4	60
75	papaya	50	50	50	6	20
88	coconut	20	10	30	5	45

Fig 4.4 Fertilizer Dataset

- **Plant Disease dataset:** 76100 items in this dataset include photos of illness leaves and 38 distinct classes. For instance, Apple->Apple scab, Apple black rot, and Apple healthy

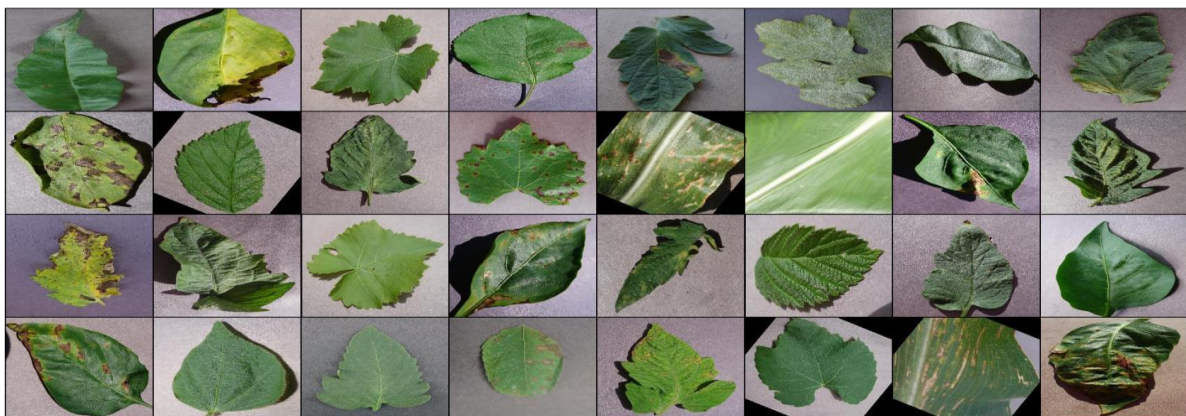


Fig 4.5 Disease Dataset

4.6 UML Diagram

Fig 4.6 visualize the way a system has been designed and structure of a system.

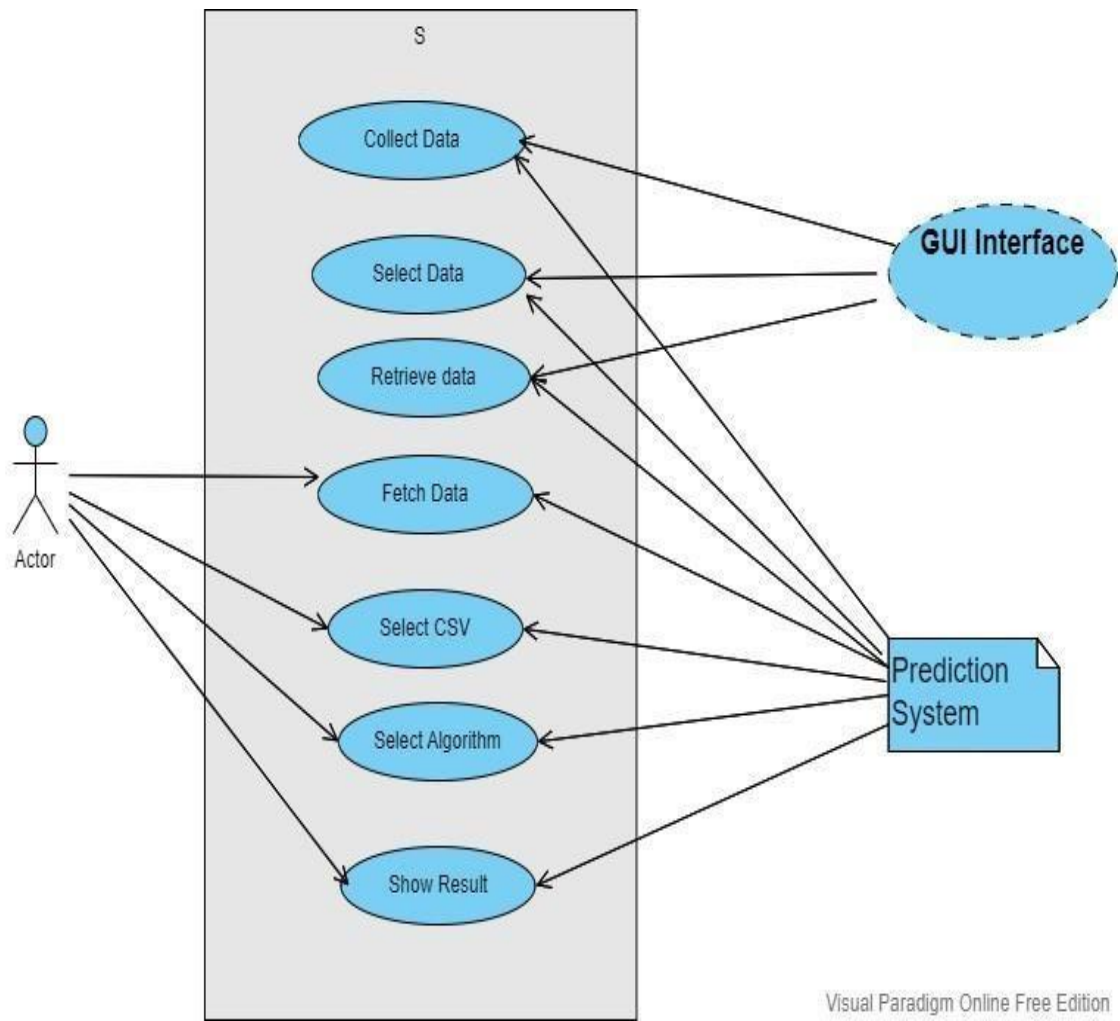


Fig 4.6 UML Diagram

4.6.1 Disease Prediction Diagram

Fig.4.7 shows the flow of the disease prediction process.

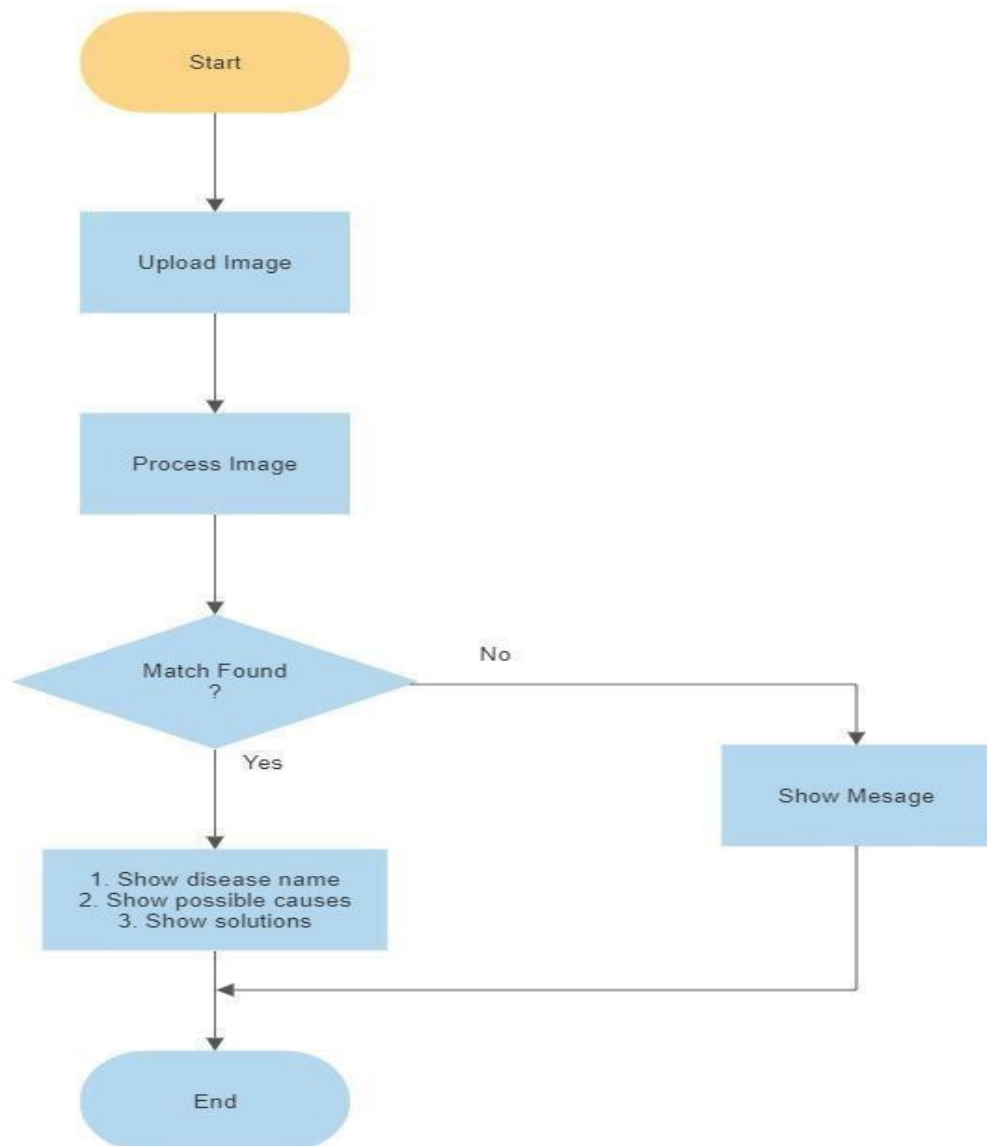


Fig. 4.7 Disease Prediction

4.6.2 Crop Recommendation Diagram

Fig 4.8 shows the flowchart of Crop Recommendation System, It includes the step by step representation of process.

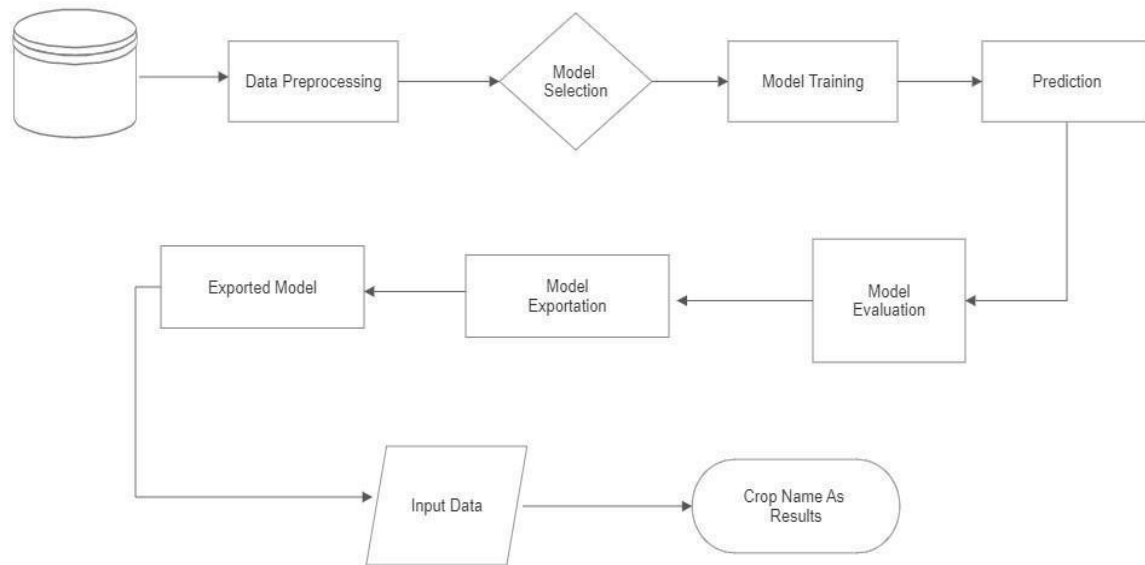


Fig. 4.8 Crop Recommendation

4.6.3 DFD Level 0 and DFD Level 1

Fig 4.4 shows data flow process of the system.

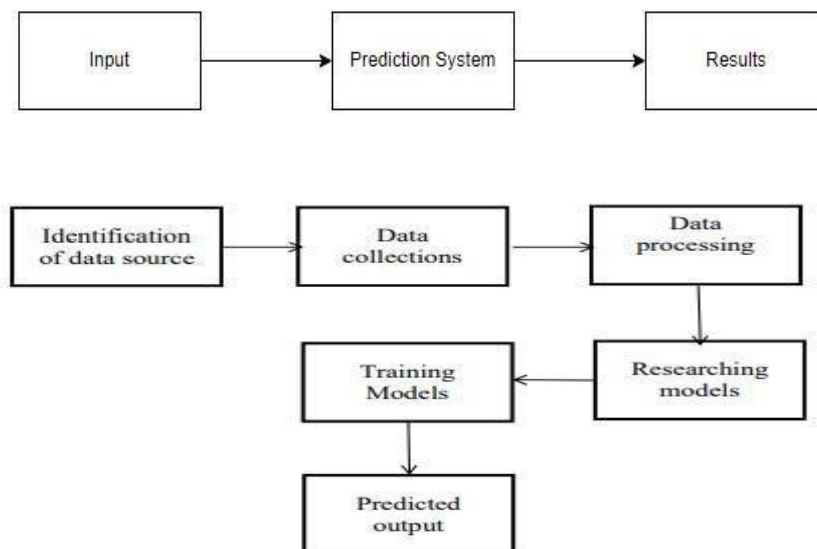


Fig. 4.9 DFD Level 0 and DFD Level 1

Chapter 5

Implementation Details

5.1. Crop Recommendation

The implementation of the crop recommendation module involves several steps, including data collection, preprocessing, feature engineering, model training, and validation. The following are the implementation details of the crop recommendation module:

1. **Data Collection:** The first step is to collect relevant data from various sources, including soil samples, weather data, crop yield data, and other environmental variables. The data should cover a broad range of factors that influence crop growth and yield.
2. **Preprocessing:** The collected data needs to be preprocessed to remove outliers, missing values, and noise. The preprocessing step also involves scaling and normalization of the data to ensure that all features have a similar range and distribution.
3. **Model Training:** The next step is to train a machine learning model using the preprocessed and engineered features. The model should be able to learn the relationship between various environmental factors and crop yield. The model can be a supervised or unsupervised learning algorithm, depending on the availability of labeled data.

5.2. Fertilizer Recommendation

The implementation of the fertilizer recommendation module involves several steps, including data collection, preprocessing, feature engineering, model training, and validation. The following are the implementation details of the fertilizer recommendation module:

1. **Data Collection:** The first step is to collect relevant data from various sources, including soil samples, weather data, crop type, and fertilizer usage. The data should cover a broad range of factors that influence fertilizer application and crop yield.

2. **Preprocessing:** The collected data needs to be preprocessed to remove outliers, missing values, and noise. The preprocessing step also involves scaling and normalization of the data to ensure that all features have a similar range and distribution.

3. **Model Training:** The next step is to train a machine learning model using the preprocessed and engineered features. The model should be able to learn the relationship between various environmental factors, crop type, and fertilizer application. The model can be a supervised or unsupervised learning algorithm, depending on the availability of labeled data.

Once the model is trained and validated, the fertilizer recommendation module can be deployed to provide farmers with accurate and personalized recommendations on the appropriate type and amount of fertilizer to apply to their crops. The module can also provide farmers with additional information on the optimal fertilizer application method, schedule, and expected yield for each recommended crop.

5.3. Disease Prediction

Disease prediction using ResNet models typically involves the following implementation steps:

1. **Data Preparation:** The first step is to collect a dataset of images that represent different diseases. This dataset can be either pre-existing or you can collect it yourself. You need to ensure that the images are of high quality and have appropriate labeling.

2. **Data Preprocessing:** Once the dataset is ready, the next step is to preprocess it. This includes resizing the images to a standard size, normalizing the pixel values, and splitting the dataset into training, validation, and testing sets.

3. **Model Architecture:** ResNet(Residual Network) is a deep learning architecture that has shown great success in image recognition tasks. There are several versions of ResNet, such as ResNet-18, ResNet-34, ResNet-50, etc. You can choose the appropriate version based on the complexity of your problem.

4. **Model Training:** The next step is to train the ResNet model using the training set. This involves feeding the images into the model and updating the weights using backpropagation. You can use a cross-entropy loss function to measure the difference between the predicted and actual labels.

5. **Model Testing:** Once you are satisfied with the performance of the model on the validation set, you can test it on the testing set. This involves measuring the same metrics as in step 5.

6. Model Deployment: Finally, you can deploy the model to predict diseases in new images. This involves preprocessing the new images and feeding them into the model to obtain the predicted label.

In summary, disease prediction using ResNet models involves data preparation, data preprocessing, model architecture, model training, model evaluation, model testing, and model deployment. Each step requires careful attention to ensure that the model performs well on the given task.

5.4 Parameters

The parameters nitrogen, phosphorus, and potassium are essential macronutrients required by plants for their growth and development. They are often referred to as NPK and are the three numbers that you commonly see on fertilizer packaging. Each nutrient has a specific role in plant growth. Nitrogen is responsible for leaf growth, phosphorus promotes root growth and potassium helps with the overall health and vigor of the plant.

Soil type is another important parameter that affects plant growth. Different types of soil have different physical and chemical properties that can affect nutrient availability, water retention, and aeration. For example, sandy soil drains quickly and can be deficient in nutrients, while clay soil retains water but can be too heavy for plants to grow well. Loamy soil is considered the ideal soil type for most plants as it has a good balance of water retention, aeration, and nutrient availability.

Humidity is the amount of water vapor present in the air. It can have a significant impact on plant growth as it affects the rate of transpiration, which is the process by which plants lose water through their leaves. High humidity can reduce the rate of transpiration, which can be beneficial for plants in dry conditions as it helps them conserve water. However, high humidity can also create conditions that are conducive to fungal diseases, which can be harmful to plants. Low humidity, on the other hand, can increase the rate of transpiration, which can be beneficial in humid conditions as it helps plants cool down and prevents fungal diseases.

In summary, nitrogen, phosphorus, and potassium are essential nutrients required for plant growth, soil type affects nutrient availability, water retention, and aeration, and humidity affects the rate of transpiration and can create conditions that are conducive to fungal diseases. All of these parameters are important to consider when growing plants as they can have a significant impact on plant health and growth.

4.4 Hardware and Software

- Hardware System Configuration:

- Processor: 2 gigahertz (GHz) or faster processor.
- RAM: 4 gigabyte (GB) for 32-bit or 4 GB for 64-bit.
- Hard disk space: => 16GB.

- Software Configuration:

- Operating System: Windows XP/7/8/8.1/10, Linux and Mac
- Coding Language: Python.
- Tools: Pandas, Numpy ,Seaborn , Pickle,Scikit-learn, Pytorch & ResNet.
- Framework: Flask.
- Other Tools : HTML, CSS, Bootstrap.

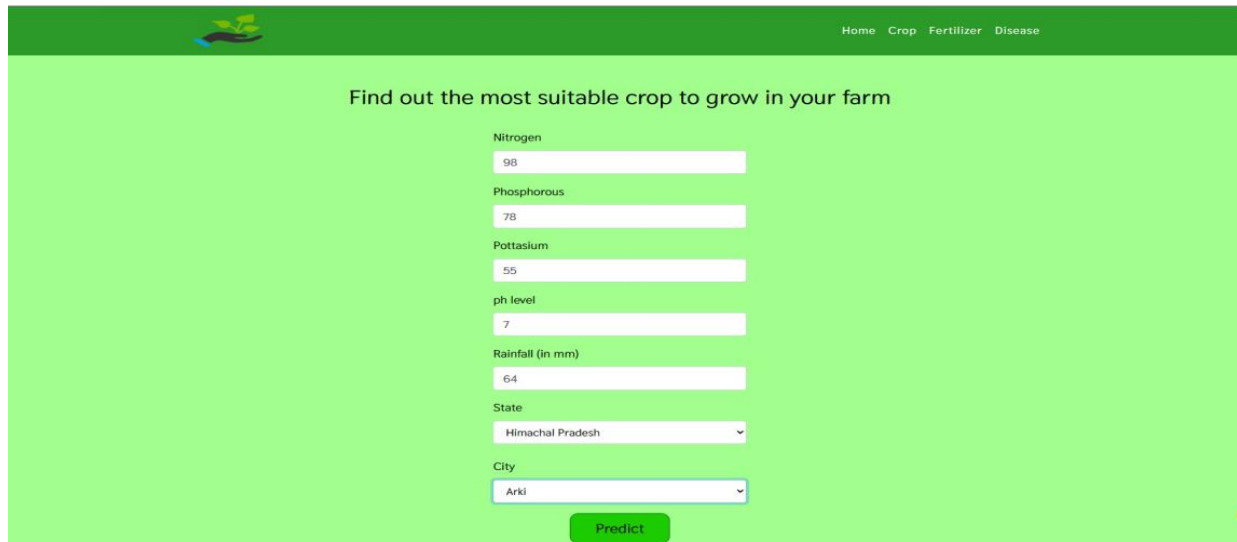
- Tools and Library used

- Library-torch, Pandas, Numpy, Matplotlib, Seaborn, Pickle, Scikitlearn, Pytorch and resnet.
- Tool- Flask

Chapter 6

Results

Crop Recommendation (Input and Output)



The screenshot shows a web application for crop recommendation. The header is green with a logo on the left and navigation links (Home, Crop, Fertilizer, Disease) on the right. The main content area has a light green background and the heading "Find out the most suitable crop to grow in your farm". Below this, there are input fields for Nitrogen (98), Phosphorous (78), Pottasium (55), ph level (7), Rainfall (in mm) (64), State (Himachal Pradesh), and City (Arki). A green "Predict" button is at the bottom.

Parameter	Value
Nitrogen	98
Phosphorous	78
Pottasium	55
ph level	7
Rainfall (in mm)	64
State	Himachal Pradesh
City	Arki


Predict

Output:



This screenshot shows the top part of the web application, including the green header with the logo and navigation links (Home, Crop, Fertilizer, Disease).

You should grow *banana* in your farm

[Home](#) [Crop](#) [Fertilizer](#) [Disease](#)

Find out the most suitable crop to grow in your farm

Nitrogen

50

Phosphorous

20

Pottasium

25

ph level

6

Rainfall (in mm)

100

State

Goa

City


Madgaon

Predict

Output:

[Home](#) [Crop](#) [Fertilizer](#) [Disease](#)

You should grow *orange* in your farm

[Home](#) [Crop](#) [Fertilizer](#) [Disease](#)

Find out the most suitable crop to grow in your farm

Nitrogen

90

Phosphorous

60

Pottasium

90

ph level

8

Rainfall (in mm)

150

State

Haryana

City


Bawal

Predict

Output:

[Home](#) [Crop](#) [Fertilizer](#) [Disease](#)

You should grow *chickpea* in your farm

Home Crop Fertilizer Disease

Find out the most suitable crop to grow in your farm

Nitrogen

32

Phosphorous

98

Pottasium

156

ph level

6

Rainfall (in mm)

87

State

Gujarat


City

Predict

Output:

Home Crop Fertilizer Disease

You should grow *chickpea* in your farm

[Home](#) [Crop](#) [Fertilizer](#) [Disease](#)

Find out the most suitable crop to grow in your farm

Nitrogen

Phosphorous

Pottasium

ph level

Rainfall (in mm)

State

Tripura

City

Khowai


Predict

Output:

[Home](#) [Crop](#) [Fertilizer](#) [Disease](#)

You should grow *maize* in your farm

Fertilizer Recommendation(Input and Output)



Home Crop Fertilizer Disease

Get informed advice on fertilizer based on soil

Nitrogen

45

Phosphorous

69

Pottasium

106

Crop you want to grow

blackgram

Predict

Output:


The N value of soil is high and might give rise to weeds.
Please consider the following suggestions:

1. *Manure* – adding manure is one of the simplest ways to amend your soil with nitrogen. Be careful as there are various types of manures with varying degrees of nitrogen.
2. *Coffee grinds* – use your morning addiction to feed your gardening habit! Coffee grinds are considered a green compost material which is rich in nitrogen. Once the grounds break down, your soil will be fed with delicious, delicious nitrogen. An added benefit to including coffee grounds to your soil is while it will compost, it will also help provide increased drainage to your soil. AZ
3. *Plant nitrogen fixing plants* – planting vegetables that are in Fabaceae family like peas, beans and soybeans have the ability to increase nitrogen in your soil
4. Plant 'green manure' crops like cabbage, corn and broccoli
5. *Use mulch (wet grass) while growing crops* - Mulch can also include sawdust and scrap soft woods

Fertilizers

If a soil test shows that the nitrogen levels are high, but the soil is deficient in other nutrients, a balanced fertilizer such as a 10-10-10 or a 14-14-14 may be used. These types of fertilizers provide a balance of nitrogen, phosphorus, and potassium, and can help to promote healthy plant growth.

Additionally, you may consider fertilizers that are specific to your crop, for example Nitrogen-Phosphorus-Potassium (NPK) fertilizers for vegetables, fruits and grains. It is always good to consult with local extension services or agronomists for best practices.


Home Crop Fertilizer Disease

Get informed advice on fertilizer based on soil

Nitrogen

Phosphorous

Pottasium

Crop you want to grow

apple

Predict

Farmer Assistant App

Helping Farmers in Agriculture
Made with ❤️ by Team CRCE

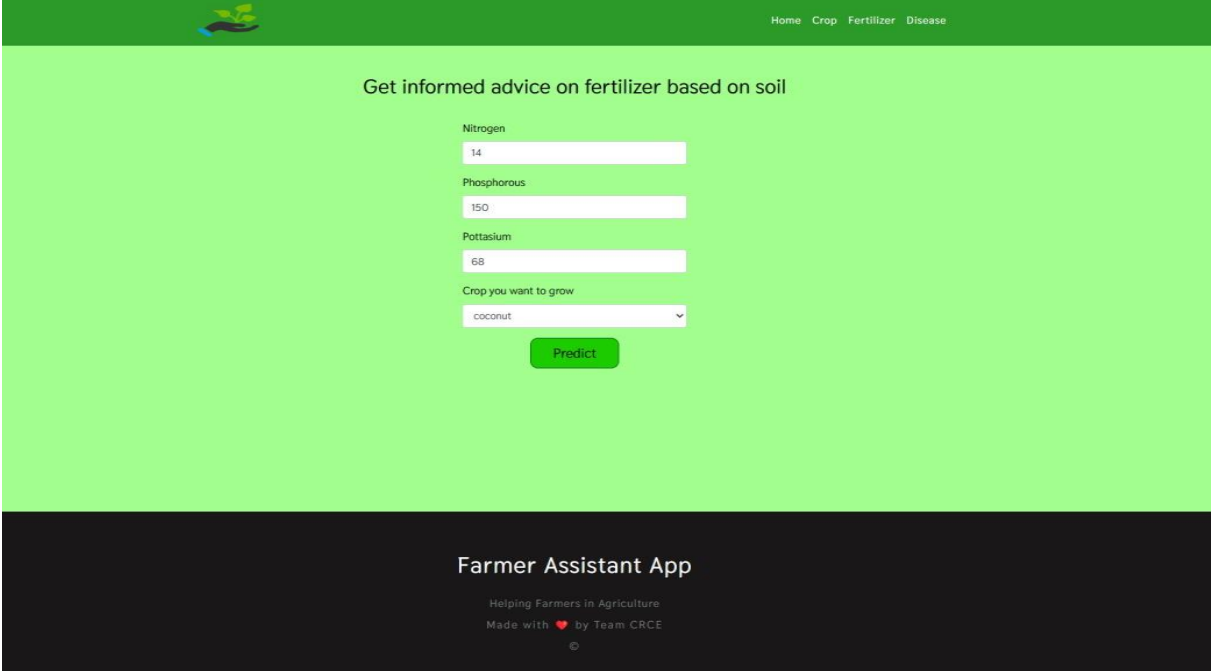
Output:

The K value of your soil is high.
Please consider the following suggestions:

1. *Loosen the soil* deeply with a shovel, and water thoroughly to dissolve water-soluble potassium. Allow the soil to fully dry, and repeat digging and watering the soil two or three more times.
2. *Sift through the soil*, and remove as many rocks as possible, using a soil sifter. Minerals occurring in rocks such as mica and feldspar slowly release potassium into the soil slowly through weathering.
3. Stop applying potassium-rich commercial fertilizer. Apply only commercial fertilizer that has a 'O' in the final number field. Commercial fertilizers use a three number system for measuring levels of nitrogen, phosphorous and potassium. The last number stands for potassium. Another option is to stop using commercial fertilizers all together and to begin using only organic matter to enrich the soil.
4. Mix crushed eggshells, crushed seashells, wood ash or soft rock phosphate to the soil to add calcium. Mix in up to 10 percent of organic compost to help amend and balance the soil.
5. Use NPK fertilizers with low K levels and organic fertilizers since they have low NPK values.
6. Grow a cover crop of legumes that will fix nitrogen in the soil. This practice will meet the soil's needs for nitrogen without increasing phosphorus or potassium.

Fertilizers

If you want to lower the potassium level in your soil, you can use a fertilizer that is lower in potassium and higher in other essential nutrients. For example, a fertilizer with a ratio of 10-10-10 (10% nitrogen, 10% phosphorus, 10% potassium) would be a good choice. You can also use fertilizers specifically formulated for your plants, such as a fertilizer for vegetables, flowers, or lawns. It's also important to use fertilizer in the correct amounts, as over-fertilizing can be harmful to plants. Always follow the instructions on the fertilizer package, and be sure to water the soil after fertilizing to help the nutrients reach the roots of your plants.



The screenshot shows the 'Farmer Assistant App' interface. At the top is a green navigation bar with a plant icon and links for 'Home', 'Crop', 'Fertilizer', and 'Disease'. The main section has a light green background with the heading 'Get informed advice on fertilizer based on soil'. Below this is a form with four input fields: 'Nitrogen' (value 14), 'Phosphorous' (value 150), 'Potassium' (value 68), and 'Crop you want to grow' (dropdown menu showing 'coconut'). A green 'Predict' button is at the bottom of the form. The footer is a dark grey bar with the text 'Farmer Assistant App', 'Helping Farmers in Agriculture', 'Made with ❤️ by Team CRCE', and a copyright symbol.

Output:

The P value of your soil is high.
Please consider the following suggestions:

1. *Avoid adding manure* – manure contains many key nutrients for your soil but typically including high levels of phosphorous. Limiting the addition of manure will help reduce phosphorus being added.
2. *Use only phosphorus-free fertilizer* – if you can limit the amount of phosphorous added to your soil, you can let the plants use the existing phosphorus while still providing other key nutrients such as Nitrogen and Potassium. Find a fertilizer with numbers such as 10-0-10, where the zero represents no phosphorous.
3. *Water your soil* – soaking your soil liberally will aid in driving phosphorous out of the soil. This is recommended as a last ditch effort.
4. Plant nitrogen fixing vegetables to increase nitrogen without increasing phosphorous (like beans and peas).
5. Use crop rotations to decrease high phosphorous levels

Fertilizers

For high phosphorous levels in the soil, a fertilizer with a low P (phosphorus) and high N (nitrogen) and K (potassium) ratio, such as a 10-10-20 or 16-4-8 fertilizer, would be a good choice. It is important to note that overuse of fertilizers can lead to soil imbalances and excess nutrients in the water supply, so it's always best to follow the manufacturer's instructions and consult with a local expert or extension service for specific recommendations for your area.



Get informed advice on fertilizer based on soil

Nitrogen

Phosphorous

Pottasium

Crop you want to grow

Predict

Output:

The P value of your soil is high.
Please consider the following suggestions:

1. *Avoid adding manure* – manure contains many key nutrients for your soil but typically including high levels of phosphorous. Limiting the addition of manure will help reduce phosphorus being added.
2. *Use only phosphorus-free fertilizer* – if you can limit the amount of phosphorous added to your soil, you can let the plants use the existing phosphorus while still providing other key nutrients such as Nitrogen and Potassium. Find a fertilizer with numbers such as 10-0-10, where the zero represents no phosphorous.
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Fertilizers

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Get informed advice on fertilizer based on soil

Nitrogen

Phosphorous

Pottasium

Crop you want to grow

Predict

The K value of your soil is high.

Please consider the following suggestions:

1. *Loosen the soil* deeply with a shovel, and water thoroughly to dissolve water-soluble potassium. Allow the soil to fully dry, and repeat digging and watering the soil two or three more times.
2. *Sift through the soil*, and remove as many rocks as possible, using a soil sifter. Minerals occurring in rocks such as mica and feldspar slowly release potassium into the soil slowly through weathering.
3. Stop applying potassium-rich commercial fertilizer. Apply only commercial fertilizer that has a '0' in the final number field. Commercial fertilizers use a three number system for measuring levels of nitrogen, phosphorous and potassium. The last number stands for potassium. Another option is to stop using commercial fertilizers all together and to begin using only organic matter to enrich the soil.
4. Mix crushed eggshells, crushed seashells, wood ash or soft rock phosphate to the soil to add calcium. Mix in up to 10 percent of organic compost to help amend and balance the soil.
5. Use NPK fertilizers with low K levels and organic fertilizers since they have low NPK values.
6. Grow a cover crop of legumes that will fix nitrogen in the soil. This practice will meet the soil's needs for nitrogen without increasing phosphorus or potassium.

Fertilizers

If you want to lower the potassium level in your soil, you can use a fertilizer that is lower in potassium and higher in other essential nutrients. For example, a fertilizer with a ratio of 10-10-10 (10% nitrogen, 10% phosphorus, 10% potassium) would be a good choice. You can also use fertilizers specifically formulated for your plants, such as a fertilizer for vegetables, flowers, or lawns. It's also important to use fertilizer in the correct amounts, as over-fertilizing can be harmful to plants. Always follow the instructions on the fertilizer package, and be sure to water the soil after fertilizing to help the nutrients reach the roots of your plants.

Disease Prediction:



Home Crop Fertilizer Disease

Find out which disease has been caught by your plant

Please Upload The Image

Oa31549c-8adb...B.Rot 0687.JPG



Output:

Crop: Grape

Disease: Black Rot

Cause of disease:

1. The black rot fungus overwinters in canes, tendrils, and leaves on the grape vine and on the ground. Mummified berries on the ground or those that are still clinging to the vines become the major infection source the following spring.
2. During rain, microscopic spores (ascospores) are shot out of numerous, black fruiting bodies (perithecia) and are carried by air currents to young, expanding leaves. In the presence of moisture, these spores germinate in 36 to 48 hours and eventually penetrate the leaves and fruit stems.
3. The infection becomes visible after 8 to 25 days. When the weather is wet, spores can be released the entire spring and summer providing continuous infection.

How to prevent/cure the disease

1. Space vines properly and choose a planting site where the vines will be exposed to full sun and good air circulation. Keep the vines off the ground and insure they are properly tied, limiting the amount of time the vines remain wet thus reducing infection.
2. Keep the fruit planting and surrounding areas free of weeds and tall grass. This practice will promote lower relative humidity and rapid drying of vines and thereby limit fungal infection.
3. Use protective fungicide sprays. Pesticides registered to protect the developing new growth include copper, captan, ferbam, mancozeb, maneb, triadimefon, and ziram. Important spraying times are as new shoots are 2 to 4 inches long, and again when they are 10 to 15 inches long, just before bloom, just after bloom, and when the fruit has set.

Pesticides

Several pesticides are available for controlling black rot on grapes. These include:

Copper-based fungicides: These fungicides, such as copper hydroxide and copper oxychloride, are effective in controlling black rot and are often used as a protective spray. They work by preventing fungal spores from germinating and infecting the vine.

Sterol biosynthesis inhibitors: These fungicides, such as propiconazole and tebuconazole, work by inhibiting the growth and reproduction of the fungus. They are usually applied as a protective spray and can be effective in controlling black rot.

QoI fungicides: These fungicides, such as pyraclostrobin, are also effective in controlling black rot and are usually applied as a protective spray.

DMI fungicides: These fungicides, such as myclobutanil and penconazole, are also effective in controlling black rot and are usually applied as a protective spray.



Find out which disease has been caught by your plant

Please Upload The Image

Choose File 00e909aa-e3a...0degFlipLR.JPG



Predict

Output:

Crop: Apple

Disease: Black Rot

Cause of disease:

Black rot is caused by the fungus *Diplodia seriata* (syn *Botryosphaeria obtusa*). The fungus can infect dead tissue as well as living trunks, branches, leaves and fruits. In wet weather, spores are released from these infections and spread by wind or splashing water. The fungus infects leaves and fruit through natural openings or minor wounds.

How to prevent/cure the disease

1. Prune out dead or diseased branches.
2. Prune out dead or diseased branches.
3. Remove infected plant material from the area.
4. Remove infected plant material from the area.
5. Be sure to remove the stumps of any apple trees you cut down. Dead stumps can be a source of spores.
6. Common fungicides used to control apple black rot include captan, thiophanate-methyl, and pyraclostrobin
7. Pesticides that can be used to control apple black rot include captan, thiophanate-methyl, and pyraclostrobin.

Find out which disease has been caught by your plant

Please Upload The Image

Oc83302d-423...RS_LB 4904.JPG



Output:

Crop: Potato

Disease: Late Blight

Late blight is a potentially devastating disease of potato, infecting leaves, stems and fruits of plants. The disease spreads quickly in fields and can result in total crop failure if untreated.

Late blight of potato was responsible for the Irish potato famine of the late 1840s.

Cause of disease:

1. Late blight is caused by the oomycete *Phytophthora infestans*. Oomycetes are fungus-like organisms also called water molds, but they are not true fungi.
2. There are many different strains of *P. infestans*. These are called clonal lineages and designated by a number code (i.e. US-23). Many clonal lineages affect both tomato and potato, but some lineages are specific to one host or the other.
3. The host range is typically limited to potato and tomato, but hairy nightshade (*Solanum physalifolium*) is a closely related weed that can readily become infected and may contribute to disease spread. Under ideal conditions, such as a greenhouse, petunia also may become infected.

How to prevent/cure the disease

1. Seed infection is unlikely on commercially prepared tomato seed or on saved seed that has been thoroughly dried.
2. Inspect tomato transplants for late blight symptoms prior to purchase and/or planting, as tomato transplants shipped from southern regions may be infected
3. If infection is found in only a few plants within a field, infected plants should be removed, disced-under, killed with herbicide or flame-killed to avoid spreading through the entire field.

Pesticides

There are several fungicides that can be used to control potato late blight caused by *Phytophthora infestans*. Some examples include:

- 1-Mefenoxam: A systemic fungicide that is taken up by the plant and provides protection against the pathogen from within.
- 2-Chlorothalonil: A broad-spectrum fungicide that can be used to control a wide range of fungal diseases, including late blight.
- 3-Mancozeb: A protectant fungicide that forms a protective barrier on the surface of the plant, preventing the pathogen from infecting the plant.
- 4-Fosetyl-Al: A systemic fungicide that is taken up by the plant and provides protection against the pathogen from within.
- 5-Cymoxanil+Mancozeb: A combination of two fungicides that provides both systemic and protectant activity.

Find out which disease has been caught by your plant

Please Upload The Image

Choose File Obe84fb6-08e....Scorch 1001.JPG



Predict

Output:

Crop: Potato

Disease: Late Blight

Late blight is a potentially devastating disease of potato, infecting leaves, stems and fruits of plants. The disease spreads quickly in fields and can result in total crop failure if untreated.

Late blight of potato was responsible for the Irish potato famine of the late 1840s.

Cause of disease:

1. Late blight is caused by the oomycete *Phytophthora infestans*. Oomycetes are fungus-like organisms also called water molds, but they are not true fungi.
2. There are many different strains of *P. infestans*. These are called clonal lineages and designated by a number code (i.e. US-23). Many clonal lineages affect both tomato and potato, but some lineages are specific to one host or the other.
3. The host range is typically limited to potato and tomato, but hairy nightshade (*Solanum physalifolium*) is a closely related weed that can readily become infected and may contribute to disease spread. Under ideal conditions, such as a greenhouse, petunia also may become infected.

How to prevent/cure the disease

1. Seed infection is unlikely on commercially prepared tomato seed or on saved seed that has been thoroughly dried.
2. Inspect tomato transplants for late blight symptoms prior to purchase and/or planting, as tomato transplants shipped from southern regions may be infected
3. If infection is found in only a few plants within a field, infected plants should be removed, disced-under, killed with herbicide or flame-killed to avoid spreading through the entire field.

Pesticides

There are several fungicides that can be used to control potato late blight caused by *Phytophthora infestans*. Some examples include:

- 1-Mefenoxam: A systemic fungicide that is taken up by the plant and provides protection against the pathogen from within.
- 2-Chlorothalonil: A broad-spectrum fungicide that can be used to control a wide range of fungal diseases, including late blight.
- 3-Mancozeb: A protectant fungicide that forms a protective barrier on the surface of the plant, preventing the pathogen from infecting the plant.
- 4-Fosetyl-Al: A systemic fungicide that is taken up by the plant and provides protection against the pathogen from within.
- 5-Cymoxanil+Mancozeb: A combination of two fungicides that provides both systemic and protectant activity.

Find out which disease has been caught by your plant

Please Upload The Image

Choose File 000ec6ea-906...OdegFlipLR.JPG



Predict

Crop: Tomato
Disease: Mosaic Virus

Cause of disease:

1. Tomato mosaic virus and tobacco mosaic virus can exist for two years in dry soil or leaf debris, but will only persist one month if soil is moist. The viruses can also survive in infected root debris in the soil for up to two years.
2. Seed can be infected and pass the virus to the plant but the disease is usually introduced and spread primarily through human activity. The virus can easily spread between plants on workers' hands, tools, and clothes with normal activities such as plant tying, removing of suckers, and harvest
3. The virus can even survive the tobacco curing process, and can spread from cigarettes and other tobacco products to plant material handled by workers after a cigarette

How to prevent/cure the disease

1. Purchase transplants only from reputable sources. Ask about the sanitation procedures they use to prevent disease.
2. Inspect transplants prior to purchase. Choose only transplants showing no clear symptoms.
3. Avoid planting in fields where tomato root debris is present, as the virus can survive long-term in roots.
4. Wash hands with soap and water before and during the handling of plants to reduce potential spread between plants.

Pesticides

1. Use of pesticides: There are no specific pesticides that target the ToMV virus. However, using pesticides to control the insect vectors that spread the virus, such as aphids and thrips, can help to reduce the spread of the disease.
2. Cultural control methods: Some cultural control methods that can help to prevent the spread of ToMV include crop rotation, removal of infected plant debris, and maintaining good hygiene in the greenhouse or field.
3. Resistant varieties: Planting tomato varieties that have been bred to be resistant to ToMV can also help to prevent the spread of the disease.

Accuracy

The four models we employed for our crop recommendation system, as shown in table no. 2.1, which have the highest accuracy.

Table No. 6.1 Crop Recommendation

Algorithm	Accuracy (%)
Logistic Regression	95
Decision Tree	90
Random Forest	99
SVM	97

In order to predict plant diseases, we now only used one model i.e the ResNet model which able to predict diseases with an accuracy of 98.03 percent.

Table no. 6.2 Disease Prediction

Algorithm	Accuracy (%)
ResNets	98.03

- Accuracy for each class for disease prediction

Table 6.3 Accuracy of each disease

	Disease Name	Accuracy
1	Apple scab	96%
2	Apple Black rot	98%
3	Cedar apple rust	97%
4	Blueberry healthy	99%
5	Cherry (including sour) healthy	95%
6	Cherry (including sour)Powdery mildew	95%
7	Corn (maize) Common rust_	96%
8	Grape Black rot	97%
9	Grape Leaf blight (Isariopsis Leaf Spot)	98%
10	Peach healthy	95%
11	Pepper, Bell Bacterial spot	96%
12	Potato Early_ light	97%
13	Potato Late blight	95%
14	Squash Powdery mildew	97%
15	Tomato Early blight	96%
16	Tomato healthy	97%
17	Tomato Late blight	95%
18	Tomato Leaf Mold	96%
19	Tomato Septoria leaf spot	97%
20	Tomato mosaic virus	96%
21	Tomato Yellow Leaf Curl Virus	96%

Chapter 7

Conclusion And Future Enhancements

7.1 Conclusion

The entire system aids in crop selection by offering information that most farmers would not be aware of, lowering the likelihood of crop failure and raising output. Additionally, it stops them from suffering losses. Millions of farmers around the nation may access the system, which can be expanded to the online. The crop recommendation system will be further developed to connect with a yield predictor, another subsystem that would also give the farmer an estimate of production if he plants the recommended crop. The outcomes are beneficial because they allow farmers to operate more effectively.

7.2 Future Enhancements

1. Integration of machine vision and deep learning: Machine vision and deep learning technologies can be used to analyze images and video data from drones, satellites, and other sources to provide highly detailed information on crop health, growth, and stress factors. By analyzing this data, the system can provide highly accurate and detailed recommendations for crop management, including fertilizer recommendations, disease prediction, and irrigation schedules.
2. Use of 5G technology: 5G technology can enable faster and more reliable data transfer between sensors, IoT devices, and other components of the crop management system. This can help to ensure real-time data collection and analysis, enabling more precise and timely recommendations.
3. Development of autonomous farming systems: Autonomous farming systems, such as robots and drones, can perform tasks such as soil sampling, planting, and crop monitoring without human intervention. These systems can collect and analyze data in real-time, allowing for rapid decision-making and response to changing conditions.
4. Integration of quantum computing: Quantum computing can provide significant advancements in the speed and accuracy of data analysis, allowing for more complex models and simulations. This can enable more accurate predictions and recommendations for crop management, including fertilizer recommendations, disease prediction, and irrigation schedules.

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Appendix – 1

Optimizing Crop Production: An Agronomic Advisor Application Based on Soil Nutrients

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ABSTRACT

Agriculture sector works as the source of raw material for non-Agricultural sectors. As much as 60% of the land is used for farming in India. It feeds around total of 1.2 billion in population. The population has also been increasing day to day and also the agriculture is not able to meet the demanded requirements for the increasing population. Crop yields, meanwhile, have already begun to suffer as a result of climate change. Unnatural climate changes can have a negative impact on food production and forecasting, which in turn can affect farmers' economics by resulting in low yields. Droughts, floods, heat waves, storms, and various more extreme weather events can destroy crops and reduce harvests, and climate change can bring about these conditions. Crop output may also be impacted by variations in temperature and rainfall patterns, which may change when crops are planted and harvested. These changes can make it difficult for farmers to predict their yields accurately, leading to economic losses. It is possible to produce precise crop prediction results by using the right parameters, such as soil nutrients properties (Nitrogen, Phosphorous, and Potassium, nutrients concentration, soil type and pH value), rainfall patterns, temperature patterns, soil structures, and other factors, such as crop diseases. A crucial aspect of agriculture is determining the best crop to grow, and in recent years, machine learning algorithms have become increasingly important in this process. This unique research is use of ML algorithms to better precisely recommend the crops based on the location.

Supervised learning classification was used for the recommendations in this study. This study's main goal is to identify the most effective feature selection and classification techniques to predict the best harvest that will thrive in a particular environment, including temperature, rainfall, and geographic location in a given state, soil properties, including phosphorus (P), potassium (K), nitrogen (N), and pH value, as well as soil type. To provide recommendations for crops that are likely to thrive in a particular environment based on the available soil nutrients, an agronomic advisor application can be developed. The application can use a suitable classification algorithm to identify the most relevant features of the soil and environment. This can help farmers optimize their crop production and maximize their yields.

Keywords: Machine Learning, Agriculture, Soil nutrients, temperature patterns, Crop Recommendation, Random Forest, SVM, Decision Tree, Logistic Regression.

1 INTRODUCTION

Choosing the appropriate crop is a critical decision for farmers since it has a significant impact on the final yield and is influenced by factors such as the environment and soil type. Selecting the right crop for a particular farm is a challenging choice that affects the yield. Expert advice on crop selection or recommendations can be time-consuming and expensive, making it difficult for many farms to afford it. Traditional methods of crop selection, such as expert consultations or field trials, can be costly and require significant investments of time and resources. As a result, many farmers may not have access to expert advice, which can limit their ability to optimize their crop production and maximize their yields. The use of machine learning algorithms and agronomic advisor applications can provide a more cost-effective and efficient alternative to traditional methods of crop selection. By analyzing relevant data and identifying the most important features, these applications can provide farmers with recommendations for the crops that are most likely to thrive in their specific environment, based on the available soil nutrients and other relevant factors.

The management of system crops to maximize agricultural productivity is one of the key areas of precision agriculture. Suggesting suitable crops based on data analysis can help increase crop production while minimizing resource usage, by identifying the most appropriate options from a dataset. These programs are crucial for decision-making because they assist users in maximizing gains or reducing risks.

It is vital to develop a system that might provide Indian farmers with predicted information so they could make informed crop decisions. In light of this, we propose a system, an intelligent system, that, before advising the user on the crop that would grow the best, would evaluate soil characteristics (N, P, K, soil type, pH value, and nutrients concentration), as well as environmental variables (rainfall, temperature, and geographic location in relation to state).

2 LITERATURE SURVEY

In paper [1], Professor Rakesh Shirsath and a number of coauthors suggested a system that helps users select the crop to be planted. The method is an online system that any registered farmer can access through a subscription and receive personalized information. The system has a module that compiles data from many sources on crops that have already been cultivated and recommends a crop that would

be a match for planting. To make the procedure easier overall, artificial neural networks are used. In case the farmer has any issues while using the system, a feedback method is offered at the end to allow the developer to make any necessary improvements.

In their research [2], Ji-chun Zhao and Jianxin Guo consider knowledge databases to be big data and make conclusions from the data. The various modules considered include users, knowledge engineers, domain experts, human-machine interfaces, inference engines, and knowledge banks. The knowledge acquisition system collects data for the decision system and creates a usable knowledge base to address the problem. The essay makes use of several Hadoop modules to extract features. It utilizes unstructured data, processes it with Hive, Mahout and NoSQL then stores the outcomes in HDFS. Only the data for the wheat crop was reported; other crops were not taken into account.

As mentioned in the paper [3], the RSF is a farming recommendation system that takes into account a data analysis, location detection module and crop growth database, storage module, and physiographic database. The related location discovery module identifies areas nearby the user's current location and researches the crops that are grown there. As a result, recommendations are offered to the user using a similarity matrix. The location detection module uses the Google API services to determine the user's current location and identify similar sites that are close by. Nevertheless, the system does not get user feedback to improve the process.

The system suggested in paper [4] by authors S. Pudumalar and associated co-authors uses an ensemble technique known as Majority Voting Technique, which taps the power of several models to improve prediction accuracy. The final forecast is accurate when the majority voting mechanism is used, even if one of the methods predicts incorrectly. KNN, Random Trees, CHAID, and Nave Bayes for ensemble are the techniques used. The key elements used in the prediction process are if-then rules. The ensemble model provided 88% accuracy.

The research by Yogesh Gandge and Sandhya [5] is a review paper that looks at several algorithms and how effective they are for use in agriculture. It was discovered that multiple linear regression offered a rice yield accuracy of 90–95%. The ID3 algorithm was used to study the decision tree and generate suggestions for the soybean crop. The third method, SVM, was applied to all the crops and had good accuracy while utilizing little computer power. A neural network was used to corn-related data to obtain 95% accuracy. Also used were LAD Tree, K-means, KNN, C4.5, J48 and Naive Bayes. The investigation came to the conclusion that the algorithms still needed to be enhanced for greater accuracy. A dataset from Kaggle.com was analyzed for a study titled Agricultural Yield Prediction using Data Mining [6]. The LAD Tree, J48, LWL, and IBK algorithms were utilized by the author to analyse the data using the WEKA tool. The accuracy was evaluated using specificity, accuracy, RMSE, mean absolute error and sensitivity. Confusion matrices were used to find the situations that each classifier correctly identified. The results suggested that pruning could lead to improved accuracy.

In their study [7] recommended employing ANN, KNN, SVM, GBDT, Random Forest, Decision Tree and Regularized Gradient Forest as seven machine-learning approaches for

crop selection. The system is designed to recover every crop that was sown as well as the timing of its growth at a particular time of the year. The crops providing the best yields are selected once the yield rate of each crop has been determined. In order to have the best yields, the approach also recommends which crops should be planted in what order.

3 METHODOLOGY

The most appropriate crop will be suggested using the proposed approach, which will make use of many soil-related characteristics. The suggested system's technique consists of multiple steps, as indicated in Fig.1.

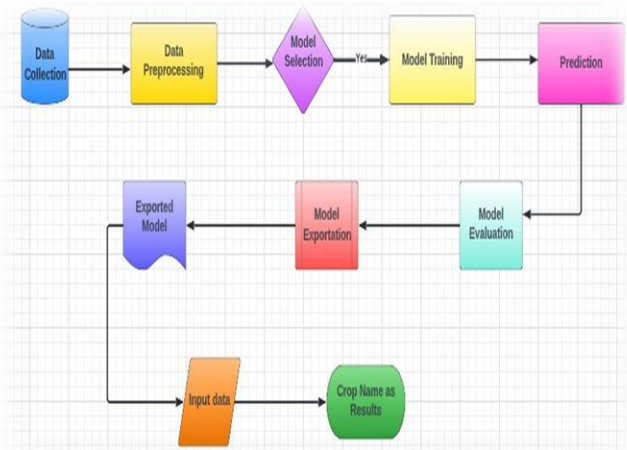


Fig.1 Proposed Architecture Diagram The

process includes:

Data collection: This involves gathering data on soil-related parameters, such as soil pH, nutrient content, and other relevant information.

Data preprocessing: The collected data is then processed and cleaned to ensure that it is accurate and suitable for analysis.

Model selection: Based on the characteristics of the problem and the data at hand, the best machine learning model is chosen.

Model training: The selected model is then trained using the preprocessed data to create a predictive model.

Prediction: After being trained, the model might be used to forecast the optimal crop based on the input data.

Model evaluation: The performance of the model is then evaluated to make sure that it is accurate and reliable.

Input data: The input data includes soil-related parameters and other relevant information that is used to generate recommendations.

Crop name as an outcome: Based on the input data, the system's final output is the suggested crop name.

Overall, this process utilizes machine learning algorithms to analyze soil-related parameters and generate personalized recommendations for suitable crops. Farmers can use the agronomic advisor application to determine what crops to grow in a particular environment.

3.1 Dataset Collection

The process of building a machine learning model begins with collecting data. It is important to collect data before constructing a machine learning model. It is crucial to obtain a sizable amount of reliable information that is pertinent to the issue at hand. Data collection allows us to keep track of past events, which allows us to use data analysis to find repeated patterns. This dataset covers characteristics specific to the soil. This primarily consists of the soil and agricultural dataset for India over the preceding 20 years. This information collection consists of over 2000 observations from over 20 distinct crops. In essence, crops rest on the ground. The soil's fertility is influenced by its texture and the presence of nutrients like nitrogen, phosphorus, and potassium. This information also includes historical weather patterns for India.

1	N	P	K	temperatu	humidity	ph	rainfall	label
2	90	42	43	20.87974	82.00274	6.502985	202.9355	rice
3	85	58	41	21.77046	80.31964	7.038096	226.6555	rice
4	60	55	44	23.00446	82.32076	7.840207	263.9642	rice
5	74	35	40	26.4911	80.15836	6.980401	242.864	rice
6	78	42	42	20.13017	81.60487	7.628473	262.7173	rice
7	69	37	42	23.05805	83.37012	7.073454	251.055	rice
8	69	55	38	22.70884	82.63941	5.700806	271.3249	rice
9	94	53	40	20.27774	82.89409	5.718627	241.9742	rice
10	89	54	38	24.51588	83.53522	6.685346	230.4462	rice
11	68	58	38	23.22397	83.03323	6.336254	221.2092	rice
12	91	53	40	26.52724	81.41754	5.386168	264.6149	rice
13	90	46	42	23.97898	81.45062	7.502834	250.0832	rice
14	78	58	44	26.8008	80.88685	5.108682	284.4365	rice
15	93	56	36	24.01498	82.05687	6.984354	185.2773	rice
16	94	50	37	25.66585	80.66385	6.94802	209.587	rice
17	60	48	39	24.28209	80.30026	7.042299	231.0863	rice
18	85	38	41	21.58712	82.78837	6.249051	276.6552	rice
19	91	35	39	23.79392	80.41818	6.97086	206.2612	rice
20	77	38	36	21.86525	80.1923	5.953933	224.555	rice

3.2 Pre-processing

Pre-processing entails taking away anomalies and disturbances from the CSV dataset. Data loss frequently occurs and has a direct impact on the final machine learning model's effectiveness and accuracy. This needs to be addressed using a variety of strategies, including median and overall column mean. We can quickly clean the dataset using sklearn. It offers an imputer class that addresses and replaces values when they are absent. The imputer class accepts arguments such as missing values and techniques the imputer employs internally. Using the mean on-axis, missing data are renewed using this method.

3.3 Feature Selection

To provide recommendations for suitable crops based on soil related parameters, the following features can be considered:

Nitrogen (N): Nitrogen is a essential nutrient that is vital for plant increase, and it performs a critical role in the improvement of plant life. The availability of nitrogen in the soil can influence the growth and yield of crops. Therefore, the nitrogen level in the soil can be a crucial feature for crop recommendation.

Phosphorus (P): Phosphorus is another crucial nutrient for plant growth and it is involved in various plant processes,

such as photosynthesis, energy transfer, and root development. The availability of phosphorus in the soil can significantly impact crop growth and yield.

Potassium (K): Potassium is a vital nutrient that is involved in several plant processes, including water regulation, photosynthesis, and disease resistance. Therefore, the availability of potassium in the soil can be an important feature for crop recommendation.

Temperature: Temperature is critical environmental aspect which could impact crop growth and improvement. Different crops have different temperature requirements for optimal growth and yield. Therefore, temperature may be a critical feature in figuring out the most appropriate crop for a particular region.

Rainfall: Rainfall is some other critical environmental element which could impact crop growth and improvement.

Different crops have different water requirements, and the availability of rainfall can significantly influence the growth and yield of crops. Therefore, rainfall can be an essential feature for crop recommendation.

Soil pH: Soil pH can have a significant impact on plant growth and development. Different crops have different pH requirements, and the availability of soil nutrients can be influenced by the soil pH. Therefore, soil pH can be a crucial feature for crop recommendation.

Humidity: Humidity is every other crucial environmental aspect which could affect crop growth and development. . Different crops have different humidity requirements for optimal growth and yield. Therefore, humidity can be a critical feature in determining the most suitable crop for a specific region.

To provide farmers with informed recommendations about what crops to grow based on their region's soil type and environmental conditions, the crop recommendation system takes into account various soil-related parameters as features. This allows farmers to make informed selections about the quality crops to develop.

3.4 Choosing Machine Learning Model

When choosing a machine learning algorithm, Random Forest is one of the most liked and widely accepted supervised learning techniques. It constitutes a number of decision trees for differential subsets of data, other than using the whole data as a single unit. This helps to enhance the accuracy of the prediction of each variable. It makes sense to use Random Forest on a large dataset as it can provide results with high accuracy in a minimum span of time. Random Forest, Decision Tree, Logistic Regression and SVM are used in this model.

3.4.1 Decision Tree

Decision trees are a type of tool that predicts outcomes based on characteristics. They use simple rules to make predictions. Decision trees can handle both numerical and nominal data.

The final prediction is made by gathering and combining all the results.

The Random Tree is a special type of decision tree that selects a random subset of characteristics to make predictions. The Random Tree builds multiple decision trees and combines their results for the best prediction

3.4.2 Random Forest

Random Forest helps to minimise overfitting in decision trees by introducing randomness in the tree construction process. The technique can handle missing data and still produce accurate results. The computational complexity of Random Forest is relatively low compared to other ensemble learning methods. Random Forest can be used for feature selection to identify the most important variables in a dataset.

The technique is insensitive to outliers and noisy data. Random Forest can be easily parallelized, Given that it is capable of handling datasets having continuous as well as categorical variables, the Random Forest is a flexible approach that may be utilized for a variety of applications. Random Forest often outperforms other machine learning algorithms when dealing with classification challenges.

The steps followed by the Random Forest algorithm are:

1. The original dataset is divided into n smaller bagged samples of size n .
2. A decision tree is constructed using input from all N bagged datasets. To calculate the ideal split, impurity measures such as Gini Impurity or Entropy are used, and M features are selected at random from the overall number of features in the training set to avoid looking at every feature in the dataset during a node split.
3. The unique outputs of each decision tree are combined into a single result.
4. For each observation, the outcomes produced by each tree are calculated, if working on a regression problem.
5. The majority vote is used to make a decision depending on the votes of the majority, if working on a classification problem.

3.4.3 Support Vector Machine (SVM)

SVM purpose is to identify a hyperplane in which the data points can be effectively separated (where N is the number of characteristics). In situations where there are two classes of data points, several hyperplanes can be utilized for classification. Data points are grouped into classes, and a hyperplane with the highest margin of difference is selected. By maximizing the margin distance, the accuracy of future data point classifications is enhanced.

3.4.4 Logistic regression

Observations are categorised into distinct classes using a method known as logistic regression. Examples of categorization problems include whether something is spam or not, whether an online transaction is fraudulent or not, and if a tumour is malignant or benign. The output of logistic regression is converted into a probability value via the logistic sigmoid function.

Which logistic regression models are there?

1. Binary

2. Failure of the class of multilinear functions, such as sheep, dogs, or cats.

An approach for predictive analysis is logistic regression, a machine learning technique that is used for classification issues and is based on the probability notion.

3.5 Crop Prediction

To determine the best crop variety for a particular region, the crop recommendation system employs a machine learning algorithm that takes into account the unique environmental factors of the area. The system utilizes user input data to train the model and identify the crop with the highest probability of success. To determine the ideal crop type, machine learning techniques including SVM, RF, logistic regression, and decision trees are used. The technology evaluates variables like humidity, soil moisture, temperature, and pH levels to suggest which crops farmers should produce.

4 RESULT ANALYSIS

The proposed crop recommendation model relies on a crop database and soil factors to suggest the best crop for a given soil type. The best crop variety is identified using machine learning algorithms, and the system found that the Random Forest technique generated the most accurate results. Table 1 shows the accuracy rates for each algorithm evaluated by the system.

Table 1 Proposed methodology performance analysis

Algorithm	Accuracy
Logistic Regression	95%
Decision Tree	90%
Random Forest	99%
SVM	97%

In this field, a number of prototypes have been put out that are helping to solve agricultural problems. Indian agriculture has enormous untapped potential. The technology that will help farmers by giving them the necessary advice on crops, their growth, and other fundamental information still has to be improved to be more compact, accurate, and affordable. The majority of the approaches now in use involve manually determining the soil type. The approach had a number of drawbacks. The system might not offer the necessary support in other circumstances. Therefore, this work suggested a novel strategy that is based on the location, to recommend crops and other strategy to address the drawbacks of existing papers.

Table 2 Existing research papers performance analysis

Algorithm Used	Accuracy	Drawback	Citation
Subscription based system, ANN	94.5%	Earlier-planted Crops unknown to the system.	[1]
Inference engine ANN	90%	Can Hadoop and Artificial Neural Networks work together to produce better results.	[2]
Recommendation generation module.	92.4%	Crop growing season, crop output rate, and physiographic factors Database of seasonal crops.	[3]
Tools like CHAID, random tree, naive bayes KNN and WEKA	88%	Missing and out of range value.	[4]
Multiple Linear Regression, SVM, Decision tree utilising ID3, K-means, C4.5, Neural Networks and KNN	MLR= 90% NN=95%	For the algorithms to become more accurate, improvement is required.	[5]
J48, LAD tree, LWL, IBK algorithm	IBK gives the highest accuracy	The LAD tree displayed the least accuracy. The tree can be pruned to reduce errors.	[6]

5. CONCLUSION

The crop recommendation system has been effective in creating a model that can foretell which crops would do best given various environmental characteristics including topography, soil type, and climate. The system analyzes vast amounts of data to identify optimal crop varieties for different regions and soil types, considering factors such as temperature, rainfall, and soil fertility. After extensive research and analysis, the system has found that specific crops are better suited for particular regions and soil types. For example, crops that require high levels of rainfall are suitable for areas with high precipitation, while crops that can tolerate drought conditions are better suited for regions with low rainfall.

The predictive model developed by the system uses machine learning algorithms to analyze the data and provide customized recommendations to farmers. To recommend the best crops for a given region, the model consider a number of variables, including soil pH, nutrient content, and other environmental circumstances. By providing personalized recommendations, the system can

help to increase crop yields, reduce costs, and improve efficiency in the agricultural sector. This, in turn, can contribute to sustainable agricultural practices and food security.

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Appendix – 2



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