Agriculture Application

Graduation Project, Part-I (CS)

Computer Science Department

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Abstract

The global food demand is projected to increase sharply by 2100. The population size is predicted to be over 9 billion people, so human society needs to increase food production by an estimated 70% by 2050 to feed an expected population. On the other hand, Infectious diseases reduce the potential yield by an average of 40% in addition to many farmers in the developing world experiencing yield losses as high as 100%. Because of these problems, a farmer is usually using the expected solution, which is hiring agricultural engineers, to help him with these problems which are costly. By the advancement of technology, we can make use of that by providing methods to help the farmers to get rid of these issues, such as using an app that can take a photo for a plant, and the application, in turn, presents you with the result of this plant, is it infected with a pest or disease, or it is healthy. This is actually it being our application, but it doesn't just do this process only, you can also ask about anything related to your plant by creating a post in the post area and get answers from other users who have a previous experience, finally, you can calculate the amount of required fertilizer based on the type of plant and land area on different periods of plant growth.

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1. Introduction

The global population has been expanding rapidly for many years, standing at around 7.3 billion in 2016, due to several factors, such as advanced maternity and healthcare so human society needs to increase food production by an estimated 70% by 2050 to feed an expected population. To feed a world population growing by up to 160 people per minute, with >90% of them in developing countries, will require an astonishing increase in food production. Unfortunately, Infectious diseases reduce the potential yield by an average of 40% with many farmers in the developing world experiencing yield losses as high as 100%. Plants act as an important resource for everyone in terms of food. If disease occurs, then it is very necessary to detect plant diseases in the early stage. There exist many models that help in detecting and classifying plant diseases. In Recent days, Machine learning is a great way to detect diseases it gives the computers the ability to learn without being explicitly programmed. There exist many models in machine learning to detect plant diseases. Some of them are K-means, KNN for classifying the leaves are healthy or diseased. Later, there is advancement in the field of machine learning which results in the evolution of deep learning. Deep learning algorithms are learning the features from input images during the training stage and exhibit results with suitable metrics. The working of deep learning is as follows in deep learning the information is passed through some layers. The output of the previous layer is given as input to the next layer. It passes the learned features from one layer to the next layers using activation functions till the output layer produces the desired outcomes. This idea can be extended for plant disease detection systems to manage and monitor wirelessly in largescale agriculture production with the use of drones for surveillance, the use

of sensors for managing the quantity of water, as well as fertilizers and light necessary for a qualitative production outcome. For this issue using our app, you can take a photo of the plant and by- processing, if the plant is infected, or not if it is then showing the type of disease and methods of organic and chemical treatment.

In addition, you can create a post according to the category you have chosen (plant type) or others. Asking for help or posting some information.

Finally, the calculator part of the amount of fertilizer you use is based on the nature of the plant and the area of the cultivated land. Plus, the best fertilization plan you can use.

2. Related Work

Through the application of artificial intelligence (AI) and machine learning (ML), growers can access increasingly sophisticated data and analytics tools, which enables better decisions, improved efficiencies, and reduced waste in food, all while minimizing negative environmental consequences.

| Name | Authors | Year | Abstract |
|-------------------|-------------|-------|--|
| An Improved | -Anjanadevi | March | In this paper, we have focused on |
| Deep Learning | Bondalapati | 2020 | plant data images in agricultural |
| model for plant | | | field. Agriculture is one of major |
| disease detection | | | living |
| | | | source in India. To increase the yield |
| | | | by preventing diseases and detection |
| | | | of diseases place major role in |
| | | | agriculture domain. By using |
| | | | Improved and customized DCNN |
| | | | model (improved-detect), We trained |
| | | | plant Doc and plant village datasets. |
| | | | Mainly we used Tomato, Corn and |
| | | | potato plant for model training and |
| | | | testing. we have experimented on |
| | | | plant image data set tomato leaves |
| | | | both healthy and diseased ones. |
| | | | Experimental results are compared |
| | | | with state of the architectures like |
| | | | Mobile Net, Dark Net-19, ResNet- |
| | | | 101and proposed model out |
| | | | PERFORMS in location and |
| | | | detection of plant diseases. |

| Using Deep | -Sharada | April 15, | Using a public dataset of 54,306 |
|---------------|-----------------|-----------|--|
| Learning for | Prasanna | 2016 | images of diseased and healthy plant |
| Image-Based | Mohanty. | | leaves collected under controlled |
| Plant Disease | -David Peter | | conditions, we train a deep |
| Detection | Hughes. | | convolutional neural network to |
| Bettetion | -Marcel Salathe | | identify 14 crop species and 26 |
| | | | diseases (or absence thereof). The |
| | | | trained model achieves an accuracy |
| | | | of 99.35% on a held-out test set, |
| | | | demonstrating the feasibility of this |
| | | | approach. When testing the model on |
| | | | a set of images collected from trusted |
| | | | online sources - i.e., taken under |
| | | | conditions different from the images |
| | | | used for training the model still |
| | | | achieves an accuracy of 31.4%. |
| | | | While this accuracy is much higher |
| | | | than the one based on random |
| | | | selection (2.6%), a more diverse set |
| | | | of training data is needed to improve |
| | | | the general accuracy. |

| Plant Disease | -Mohamet Faye. | January | The combination of high-end smart- |
|------------------|----------------|---------|---------------------------------------|
| Detection with | -Chen Bingcai. | 2020 | phones and computer vision via Deep |
| Deep Learning | -Kane Amath | | Learning has made possible what can |
| and Feature | Sada | | be defined as "smartphone-assisted |
| Extraction Using | | | disease diagnosis". In the area of |
| Plant Village | | | Deep Learning, multiple architecture |
| | | | models have been trained, some |
| | | | achieving performance reaching |
| | | | more than 99.53% . In this study, we |
| | | | evaluate CNN's architectures |
| | | | applying transfer learning and deep |
| | | | feature extraction. All the features |
| | | | obtained will also be classified by |
| | | | SVM and KNN. Our work is feasible |
| | | | by the use of the open-source Plant |
| | | | Village Data set. The result obtained |
| | | | shows that SVM is the best classifier |
| | | | for leaf disease detection. |

| Plant Disease | -Shima Ramesh | April 2018 | we make use of Random Forest in |
|------------------|---------------|------------|---|
| Detection Using | Maniyath. | 1 | identifying between healthy and |
| Machine Learning | -Vinod P V | | diseased leaf from the data sets |
| 8 | | | created. Our proposed paper includes |
| | | | various phases of implementation |
| | | | namely dataset creation, feature |
| | | | extraction, training the classifier and |
| | | | classification. The created datasets of |
| | | | diseased and healthy leaves are |
| | | | collectively trained under Random |
| | | | Forest to classify the diseased and |
| | | | healthy images. For extracting |
| | | | features of an image, we use |
| | | | Histogram of an Oriented Gradient |
| | | | (HOG). Overall, using machine |
| | | | learning to train the large data sets |
| | | | available publicly gives us a clear |
| | | | way to detect the disease present in |
| | | | plants in a colossal scale |

| Dlant Dissess | -P. Prathusha. | July 2020 | Machine learning is a trending area |
|------------------|------------------|-----------|--|
| Plant Disease | -K. E. Srinivasa | | where the technological benefits can |
| Detection Using | -K. Srinivas | | be imparted to the agriculture field |
| Machine Learning | | | also. It is rather inexpensive to detect |
| Algorithms | | | the diseases in plants using machine |
| | | | learning techniques rather than using |
| | | | chemical pesticides. This paper |
| | | | makes a review on the existing |
| | | | techniques and also suggests the best |
| | | | technique which can be implemented |
| | | | by farmers to recognize the disease |
| | | | faster and which proves to be |
| | | | economical to them. In this work we |
| | | | use KNN algorithm which is one of |
| | | | the best machine learning |
| | | | algorithms. |
| | | | we present a Deep Learning approach |
| Plant Disease | -Aravindhan | August | to detect and classify plant diseases |
| Detection and | Venkataramanan. | 2019 | by examining the leaf of a given |
| Classification | -Pooja Agarwal | | |
| Using Deep | | | plant. The classification is performed |
| Neural Networks | | | in multiple stages to eliminate |
| | | | possibilities at every stage, hence |
| | | | providing better accuracy during |
| | | | predictions. A YOLOv3 object |
| | | | detector is used to extract a leaf from |
| | | | the input image. The extracted leaf is |
| | | | analysed through a series of |
| | | | ResNet18 models. These ResNet18 |
| | | | models were trained using transfer |
| | | | learning. One layer identifies the type |
| | | | of leaf and the following layer checks |
| | | | for the possible diseases that could |
| | | | occur in the plant. |

| Plant disease | G.Saradhambl. | January | We propose an enhanced k-mean |
|-------------------|---------------|---------|--|
| detection and its | R. Dhivya. | 2018 | clustering algorithm to predict the |
| solution using | S. Latha. | | infected area of the leaves. A colour- |
| image | R. Rajesh. | | based segmentation model is defined |
| classification | | | to segment the infected region and |
| | | | placing it to its relevant classes. |
| | | | Experimental analyses were done on |
| | | | samples images in terms of time |
| | | | complexity and the area of infected |
| | | | region. Our project is used to detect |
| | | | the plant diseases and provide |
| | | | solutions to recover from the disease. |
| | | | It shows the affected part of the leaf |
| | | | in percentage. We planned to design |
| | | | our project with voice navigation |
| | | | system, so a person with lesser |
| | | | expertise in software should also be |
| | | | able to use it easily. |

| Imaga Dagad | -Rehanullah | June 2021 | The technology used in medical |
|--------------------|---------------|-----------|--|
| Image-Based | Khan. | | procedures has not been adequate to |
| Detection of Plant | -Khalil Khan. | | detect all diseases on time, and that is |
| Diseases: From | -Waleed | | why some diseases turn out to |
| Classical Machine | | | |
| Learning to Deep | Albattah. | | become pandemics because they are |
| Learning Journey | -Ali Mustafa | | hard to detect on time. Our focus is to |
| | Qamar | | clarify the details about the diseases |
| | | | and how to detect them promptly |
| | | | with artificial intelligence. We |
| | | | discuss the use of machine learning |
| | | | and deep learning to detect diseases |
| | | | in plants automatically. Our study |
| | | | also focuses on how machine |
| | | | learning methods have been moved |
| | | | from conventional machine learning |
| | | | to deep learning in the last five years. |
| | | | Furthermore, different data sets |
| | | | related to plant diseases are discussed |
| | | | in detail. The challenges and |
| | | | problems associated with the existing |
| | | | systems are also presented. |

| Deep learning | Konstantinos | February | convolutional neural network models |
|-------------------|--------------|----------|--|
| models for plant | Ferentinos | 2018 | were developed to perform plant |
| disease detection | | | disease detection and diagnosis using |
| and diagnosis | | | simple leaves images of healthy and |
| unu unugnosis | | | diseased plants, through deep |
| | | | learning methodologies. Training of |
| | | | the models was performed with the |
| | | | use of an open database of 87,848 |
| | | | images, containing 25 different plants |
| | | | in a set of 58 distinct classes of |
| | | | [plant, disease] combinations, |
| | | | including healthy plants. Several |
| | | | model architectures were trained, |
| | | | with the best performance reaching a |
| | | | 99.53% success rate in identifying |
| | | | the corresponding [plant, disease] |
| | | | combination (or healthy plant). |
| 1 | | I | 1 |

3. Background

The ratio of food demand overproduction is the main factor affecting the level of food security. The global population has been expanding rapidly for many years, standing at around 7.3 billion in 2016, due to a number of factors, such as advanced maternity and healthcare. However, the rise brings with it several challenges around global sustainability, including the need for more food. Agriculture contributes to 6.4% of the entire world's economic production. In at least nine countries of the world, agriculture is the dominant sector of the economy. Agriculture not only provides the fuel for billions of people but also employment opportunities for many people. The agricultural industries are seeking innovative approaches for improving crop yielding because of unpredictable climatic changes, the rapid increase in population growth, and food security concerns. Thus, artificial intelligence in agriculture also called "Agriculture Intelligence" is progressively emerging as a part of the industry's technological revolution.

The spread of agricultural diseases that destroy all agricultural crops and lack of arable land are the most problem which faces our planet, today's so we try by our project to use modern solutions using internet and technologies to beat these problems and increasing the productivity of the crops and trying to help farmers and providing them with the necessary experience to identify diseases before they occur and giving them suggestions to ensure the best productivity. ML-driven image processing allows farmers to rely upon digital tools to recognize weed species and to determine which crops are healthy and which ones are infested with disease caused by fungi, bacteria, or viruses.

We focus on using machine learning algorithms and computer vision to build models which have the ability to detect diseases and give users the best solution to its problem and try also to connect experienced agricultural engineers with simple farmers which helps to decrease time and cost to anyone who needs the advice.

4. Domain Analysis

The widespread distribution of smartphones among crop growers around the world with an expected 5 billion smartphones by 2020 offers the potential of turning the smartphone into a valuable tool for diverse communities growing food. One potential application is the development of mobile disease diagnostics through machine learning and crowdsourcing.

The main issue is: trying to help farmers know the type of pests that affect their crops, in addition to knowing the appropriate treatment for this pest as soon as possible. Some new farmers also suffer from lack of experience in some agricultural matters, so they need someone to exchange experience within order to have the best crop. It can be obtained in these conditions and using the appropriate tools and equipment available.

Identification of the pest: The farmer will be able to take a picture with his smart phone of the leaf of the plant that he wants to know the type of pest infecting him. And then he uploads this image to the application on his phone so that the application analyzes the image and gives him the results of his analysis of the image and gives him the best solutions and suggestions to treat this scourge.

Applying Machine learning algorithms and classifiers, they showed promising results in image classification and decision-making to assist growers in their diagnosis, which will serve as a boost in improving plant care services through effective image analysis of symptoms (pests) suffered by the plant.

Helping others/asking for help: Any farmer or anyone who wants opinion or even advice on his plant or anything related to the plant he is interested in can ask a general question to all present or users of the application and one of them will respond to him, which increases him with general information, or it saves him the trouble of searching in vain.

Fertilization: Calculating the appropriate amount of fertilizer for the cultivated land; some plants suffer from a lack or increase in the percentage of fertilizer used, so the application will provide the correct standards that you can use in the cultivated land area to obtain the best fertilization.

5. Techniques

ML Model: Applying Machine learning algorithms and classifiers, they showed promising results in image classification and decision making to assist growers in their diagnosis, which will serve as a boost in improving plant care services through effective image analysis of symptoms (pests) suffered by the plant.

Application components:

- **❖ Login page:** This page has the username and password, and it also has a button for the registration page if the user is not registered.
- ❖ **Registration page:** A page that enables the user to register, to be able to use the application.
- ❖ Plant Fertilization Page: Every plant has certain conditions in its composting, and that's why fertilizing a plant is different from the other and that the app will provide you that you know the appropriate amounts of fertilizer in relation to the area of the land for each plant.
- ❖ Posts Page: On this page, each user will be able to view other people's questions, advice, and responses to them, according to the latest and closest to you in terms of your geographical location, so that you can find environmental content that is close to you in the environment, climate, plants and crops that you are related to, so that it will be more useful to you. He can also download a question or inquiry that will appear to all users of the application; If you want to search for specific questions or inquiries related to a particular plant, you can search by the tag that will be available in any question or consultation.

6. Risks and Constrains

> Risks

| Risk | Strategy | Priority |
|--------------------------|---------------------|----------|
| User may not know | Show electronic | High |
| how to use the | pictures to how use | |
| application | the application at | |
| | the first use. | |
| Malfunction with the | Hosting the | High |
| servers that host the | database and the | |
| database, the posts, and | several servers. | |
| users' information | | |
| User may scan an | We try to show | High |
| untrained picture or | him the related | |
| unclear one to the | details about his | |
| application | picture. | |
| Malfunction with the | Hosting the model | High |
| server or the API | and the API on | |
| hosting the Machine | several servers. | |
| Learning Model | | |
| Internet for signed user | We will show him | High |
| may be failed | the saved posts and | |
| | details about | |
| | pictures he scanned | |
| | them before. | |

> Constrains

- ➤ The application should provide friendly and easy interface with not many icons and images to help people who don't know use the app directly.
- ➤ User should scan objects clearly as possible
- ➤ User should have internet to deal with all application features.
- ➤ User should have the application on his device
- ➤ Maintenance of an end-to-end encryption mechanism for providing confidentiality is a must.

7. Project Plan

| phase | Oct | Nov | Dec | Jan | Feb | Mar | Apr | May | Jun |
|----------------|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| Gathering | | | | | | | | | |
| Information | | | | | | | | | |
| Define | | | | | | | | | |
| Requirements | | | | | | | | | |
| analysis | | | | | | | | | |
| design | | | | | | | | | |
| implementation | | | | | | | | | |
| Develop ML | | | | | | | | | |
| Model System | | | | | | | | | |
| Testing and | | | | | | | | | |
| Final | | | | | | | | | |
| Discussion | | | | | | | | | |

➤ Quality Assurance Plan

Black box: In this stage we use test dataset as input to our ML model to ensure the accuracy of output of the system.

White box:

Unity Testing: - In this stage of testing, we will take every component of our system such as web service, machine learning model, android application to test them separately.

Integration Testing: - In this stage of testing, we will take every component of our system such as web service, machine learning model, android application to test them separately.

Validation Testing: - Validation testing is the process of ensuring if the tested and developed application satisfies its functionality requirements. The business requirement logic or scenarios must be tested in detail. All the critical functionalities of an application must be tested here.

Alpha: - In this part, a group of testers in our team test the product in a laboratory environment to ensure efficiency of product and fix errors.

Beta: - At this stage we will test the application on real users, farmers and people who have some plants at home, retrieve feedback to our team.

8. Requirements

8.1 Functional Requirements

- 1 A user interface that enables users create accounts, add their information, the planets they are interested and the location of the users.
- 2 Verification email is sent to user whenever he/she registers for the first time on some software system.
- 3 The application allows to user to enter password and username to login to app.
- 4 Authentication of a user when he/she tries to log into the system.
- 5 The application must store all Users data
- 6 The application shows to the user all posts of plants who is interested in.
- 7 Machine learning model to allow to user capture / upload the plant image for plant type detection and classification. The model is also used for plant examination from pests and diseases, if plant is diseased, it returns details about this disease and recommends Treatments.
- 8 A Server that stores users data and Machine Learning Models.
- 9 The user enters the plant type and agricultural area and the application outputs the best time plan and quantity of fertilizer to grow well.

8.2 Nonfunctional requirements

- 1 Emails should be sent with a latency of no greater than 12 hours.
- 2 The application shows the post by the order of the tag plants and the location of the users.
- 3 User can search for posts with plant tag.

1 - Performance requirement

- The application must respond any operation in less than 4 seconds.

2 - Safety and security requirements

- The application must not affect, harm, or damage users and their mobiles.

3 - Availability

- The application must be available within 24 hours every day.

4 - Usability

The application allows to users interact with the application to achieve required goals

effectively and efficiently.

9. System Request

The application helps framers to improve the quality of the plants by providing them good care, fertilizer and tips. It also helps farmers specially who grow a small area of the land or who grow certain plants at home to save money from consulting agricultural engineers, which makes it a good saver of money for those who do not have the financial ability to consult. The application is also a time saver, because it contributes to obtaining consultations quickly through the interaction of specialists on the post. This makes the application very important in critical times for plants that need quick consultations before they die.

> Functionality

- 1- The Ability to Examinate plants and treatment for free.
- 2 The speed of obtaining the plants diseases.
- 3 The ability to generate full reports about plant diseases and treatments.
- 4 The ability to share benefits by other users.
- 5 The ability to get best time plan and best fertilizer for the plant to get a good yield.

> Expected Value

- 1- Less-cost plants examinations process.
- 2 Lost cost plant cultivation plan.
- 3- More organized posts about benefits of plants.
- 4- Ease-to-use application and friendly User-interfaces.
- 5- Increase productivity.
- 6- Customer-satisfaction.

10. System Use-Case

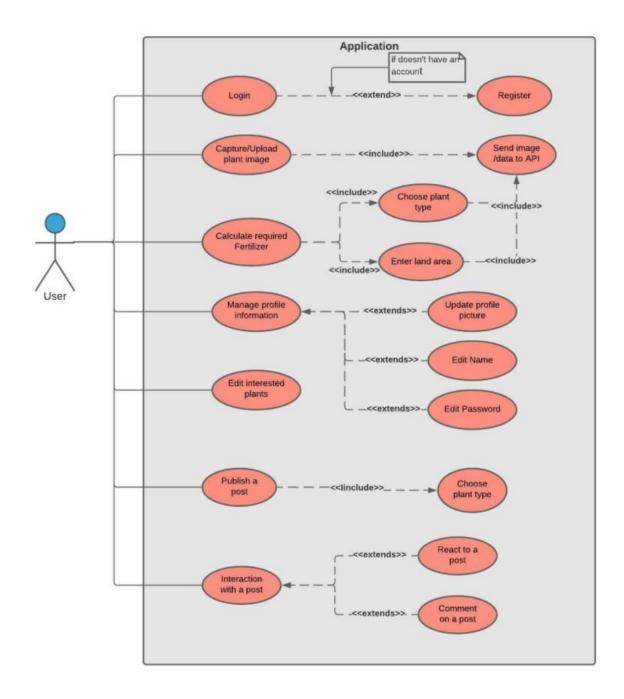


Figure 1. Application Use case

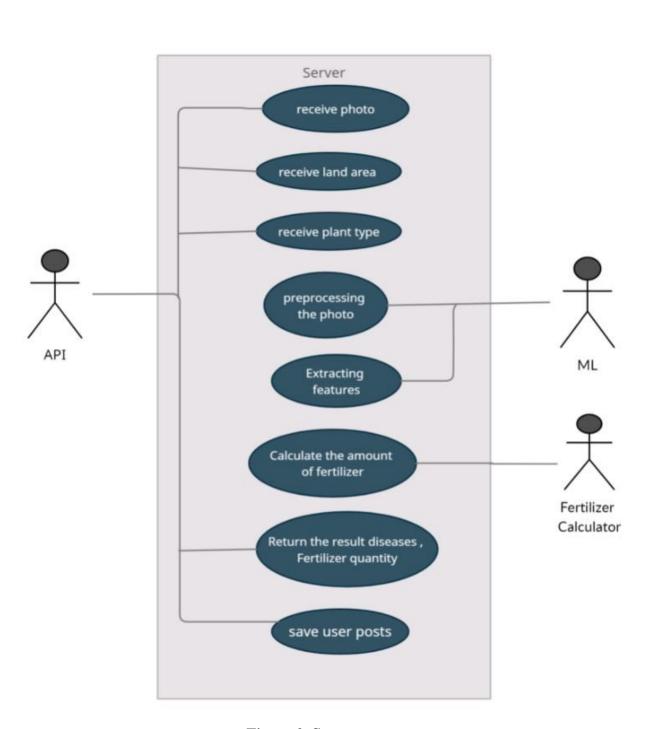


Figure 2. Server use case

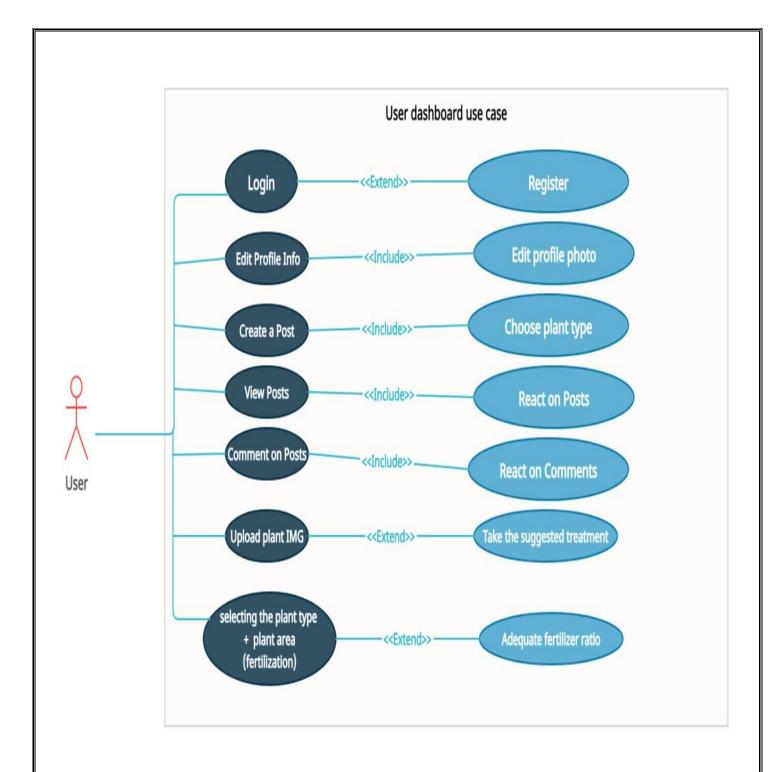
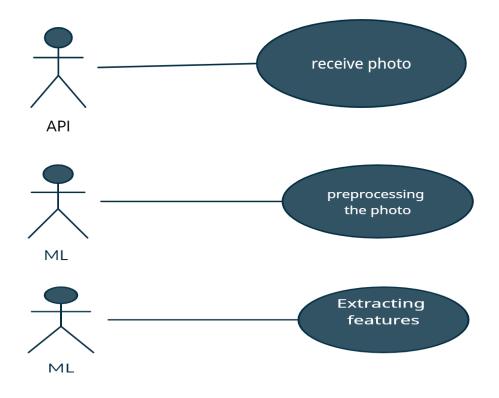
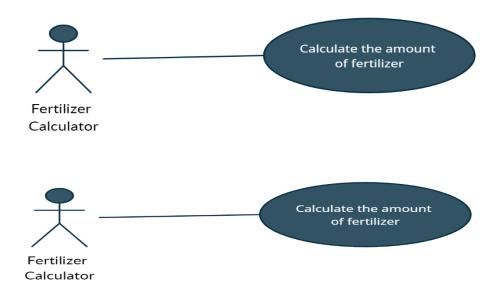


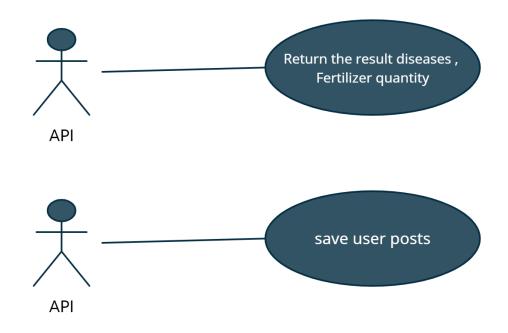
Figure 3. User dashboard use case



| Use Case ID: | ML01 | | | |
|------------------------|--|--|--|--|
| Use Case Name: | Preprocessing the photo and Extract features | | | |
| Area: | Server | | | |
| Actor: | ML | | | |
| Description: | Extract the features from received photo | | | |
| Preconditions: | A valid photo | | | |
| Postconditions: | ML module has successfully extracted features from | | | |
| | received photo | | | |
| Triggering Event: | API send a photo | | | |
| Main Flow: | 1. Receive the photo from API | | | |
| | 2. Preprocess the photo | | | |
| | 3. Extract the features from the photo | | | |
| | 4. Return the result of the health status of the plant | | | |
| Alternative Flow: | If user upload invalid photo, warning message should | | | |
| | appear to the user to reupload the data. | | | |
| | If the plant not included with our plants, warning | | | |
| | message should appear to the user that app couldn't | | | |
| | check the data. | | | |

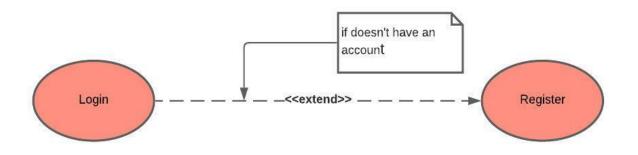


| Use Case ID: | Fertilizer_Calc01 |
|--------------------------|--|
| Use Case Name: | Calculate the amount of fertilizer. |
| Area: | Server |
| Actor: | Fertilizer Calculator |
| Description: | Getting the land area and the plant then type return the |
| | amount of fertilizer |
| Preconditions: | There are no preconditions. |
| Postconditions: | Return the amount of the fertilizer. |
| Triggering Event: | API send the land area and plant type to Fertilizer |
| | Calculator. |
| Main Flow: | 1. Received the land area. |
| | 2. Received the plant type. |
| | 3. Calculate the fertilizer amount. |
| | 4. Return the amount of the fertilizer amount. |
| Alternative Flow: | If the land area not valid number, a warning message |
| | should appear to the user. |



| Use Case ID: | API01 |
|--------------------------|--|
| Use Case Name: | Return the result and save posts. |
| Area: | Server |
| Actor: | API |
| Description: | API save user posts and return the result diseases, |
| | Fertilizer quantity. |
| Preconditions: | There are no preconditions. |
| Postconditions: | API return the result of plant status. |
| Triggering Event: | Received a photo and plant type or land area. |
| Main Flow: | Users take or upload a plant photo. |
| | API send a photo to ML module. |
| | Users enter land area and plant type. |
| | API send land area to Fertilizer Calculator |
| Alternative Flow: | If the land area not valid number, a warning message |
| | should appear to the user. |
| | If user upload invalid photo, warning message should |
| | appear to the user to reupload the data. |

Register



| Use Case ID: | Zar3a001 |
|--------------------------|--|
| Use Case Name: | Register |
| Area: | Application |
| Actor: | User |
| Description: | User creates an account |
| Preconditions: | User needs to download the application |
| | User needs an internet access |
| Post conditions: | User has successfully created an account |
| Triggering Event: | User clicks "Register" button. |
| Main Flow: | 1. Open the application |
| | 2. User enters his data |
| | 3. Users submit his data to system |
| | 4. The system validates the user data |
| | 5. The system sends a verification mail to the user |
| Additional info for | Step 2: data is name, username, password, email. |
| steps: | |
| Alternative Flow: | -if user enter non-valid data, an error message should |
| | appear to him. |

• Login



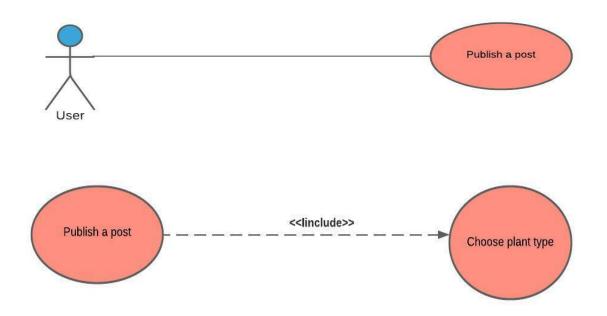
| Use Case ID: | Zar3a002 |
|--------------------------|---|
| Use Case Name: | Login |
| Area: | Application |
| Actor: | User |
| Description: | User login to the application using it's account |
| Preconditions: | User must download the application |
| | User must an internet access |
| | User must have an account |
| Post conditions: | User has successfully logged into the application |
| | The system displays the main page of application |
| | |
| Triggering Event: | User clicks "Login" button. |
| Main Flow: | 1. Open the application |
| | 2. User enters his data |
| | 3. Users submit his data to system |
| | 4. The system validates the user data |
| Additional info for | Step 2: data is username and password. |
| steps: | * |
| Alternative Flow: | If user enter non-valid data, an error message should |
| | appear to him. |

• Edit interested plants



| Use Case ID: | Zar3a003 |
|--------------------------|--|
| Use Case Name: | Edit interested plants |
| Area: | Application |
| Actor: | User |
| Description: | Users edit the pre-selected interested plants. |
| Preconditions: | User must download the application |
| | User must an internet access |
| | User must have an account |
| | User must login to the application |
| Post conditions: | Application updates the selected interested plants |
| | to the user. |
| Triggering Event: | Users click "Update" button. |
| Main Flow: | User login to the application |
| | Users select the items |
| | Users submit the items |
| | The application validates the items |
| | The system updates the selected items |
| Additional info for | Step 2, 3, 4, and 5: items are interested plants. |
| steps: | |
| Alternative Flow: | If user doesn't select any item, an error message will be |
| | appeared. |

• Publish a post



| Use Case ID: | Zar3a004 |
|--------------------------|---|
| Use Case Name: | Publish a post |
| Area: | Application |
| Actor: | User |
| Description: | Users publish a post in the community area to get help in specific plant. |
| Preconditions: | User must download the application User must an internet access User must have an account User must login to the application |
| Post conditions: | Application will make the post available to all users of the application |
| Triggering Event: | Users click "Post" button. |
| Main Flow | User login to the application Users choose plant type Users submit the post The system validates the post |
| Alternative Flow: | If user doesn't choose a plant type or doesn't write any text in the post, an error message will be appeared. |

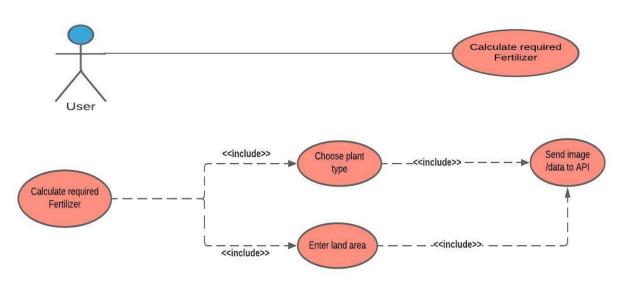
• Capture/Upload plant image





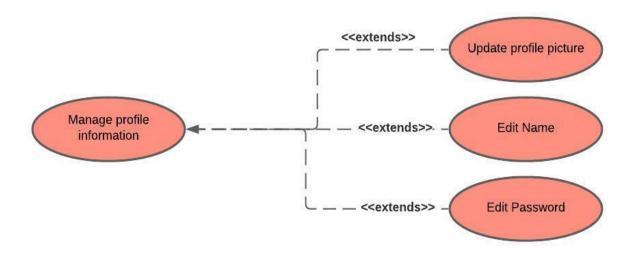
| Use Case ID: | Zar3a005 |
|--------------------------|---|
| Use Case Name: | Capture/Upload plant image |
| Area: | Application |
| Actor: | User |
| Description: | User upload or take an image of plant to see if it has a |
| | disease or pest. |
| Preconditions: | User must download the application |
| | User must an internet access |
| | User must have an account |
| | User must login to the application |
| Post conditions: | The User will have a paragraph shows the result. |
| Triggering Event: | Users click "Take Photo" button. |
| Main Flow | 1. User login to the application |
| | 2. User upload or take a photo |
| | 5. Users submit the photo |
| | 6. The system validates the photo |
| | 7. The photo is sent to the API. |
| Alternative Flow: | If user doesn't upload a plant image the application will |
| | display an error message. |

• Calculate required Fertilizer



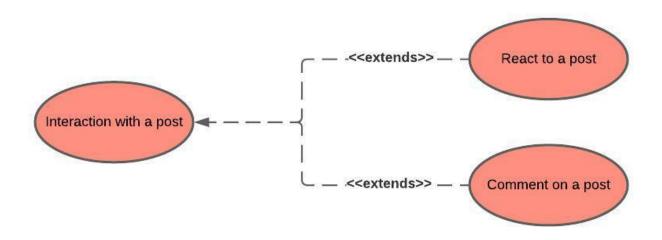
| Use Case ID: | Zar3a006 |
|--------------------------|--|
| Use Case Name: | Calculate required Fertilizer |
| Area: | Application |
| Actor: | User |
| Description: | User wants to know the amount of fertilizer suitable to |
| | his land area. |
| Preconditions: | User must download the application |
| | User must an internet access |
| | User must have an account |
| | User must login to the application |
| Post conditions: | User will have the required amount of fertilizer. |
| Triggering Event: | Users click "Calculate" button. |
| Main Flow | 1. User login to the application |
| | 2. Users enter data |
| | 3. Users submit the data |
| | 4. The system validates the data |
| | 5. The data is sent to the API. |
| Additional info for | Step 2, 3, 4, and 5: data is the plant type and the land |
| steps: | area. |
| Alternative Flow: | If user enters invalid data, an error message will be |
| | appeared. |

• Manage Profile information:



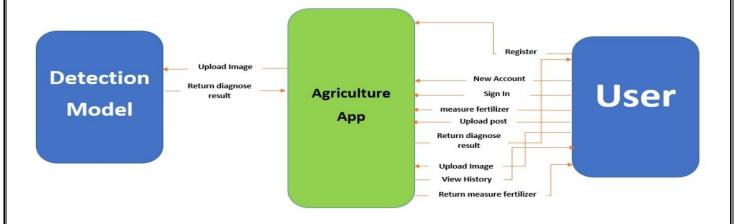
| Use Case ID: | Zar3a007 | |
|----------------------------|---|--|
| Use Case Name: | User Information | |
| Area: | Application | |
| Actor: | User | |
| Description: | Manage Profile Info | |
| Preconditions: | User must have an account on the application | |
| | User must be logged into the application successfully. | |
| Post conditions: | User has successfully logged to his/her account | |
| Triggering Event: | User clicks "Update profile data" button. | |
| Main Flow: | Open the application User enters his data Users submit his data and photo to the system The system validates the user data | |
| Additional info for steps: | Step 2: data is name, profile image, password. | |
| Alternative Flow: | If user enter non-valid data, an error message should appear to him. Ex: If the user uploads a file in the place designated for uploading a profile photo. | |

• Interaction With a post:



| Use Case ID: | Zar3a008 |
|----------------------------|--|
| Use Case Name: | Posts actions |
| Area: | Application |
| Actor: | User |
| Description: | Interaction with a post |
| Preconditions: | User must have an account on the application |
| | User must be logged into the application successfully. |
| Post conditions: | User has successfully logged to his/her account |
| Triggering Event: | User clicks "react" icon. |
| Main Flow: | 5. Open the application |
| | 6. User enters his data |
| | 7. Users submit his data and photo to the system |
| | 8. The system validates the user data |
| Additional info for steps: | User can react on any post. Step 2: |

11. Analysis Class



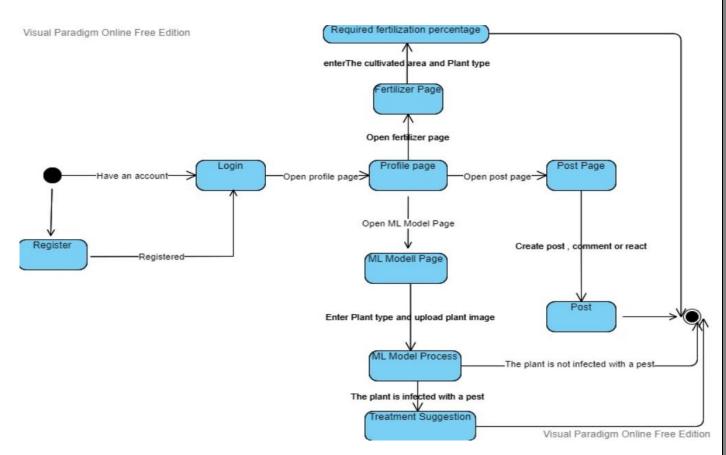


Figure 4. Context diagram

Figure 5. State diagram

12. Design Class

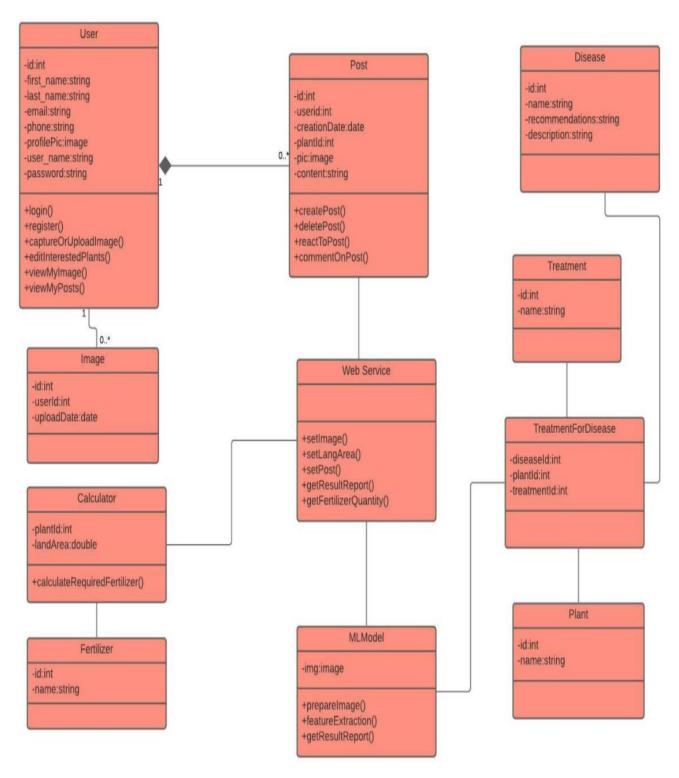
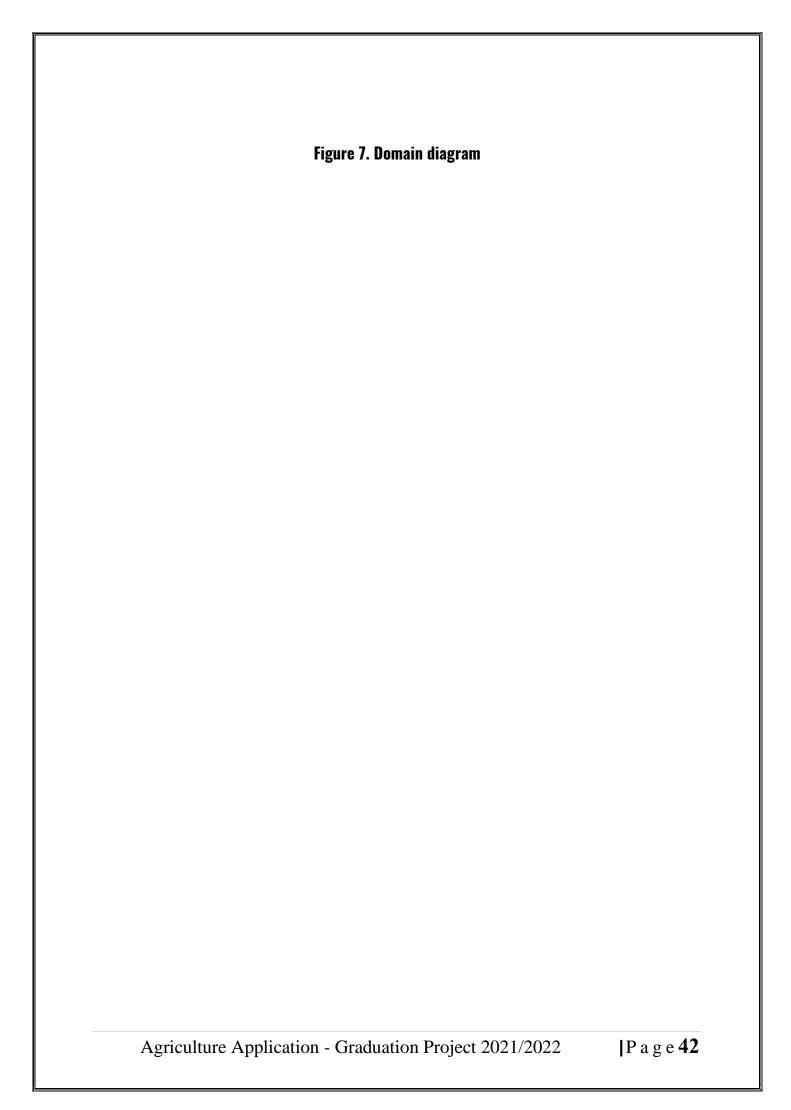
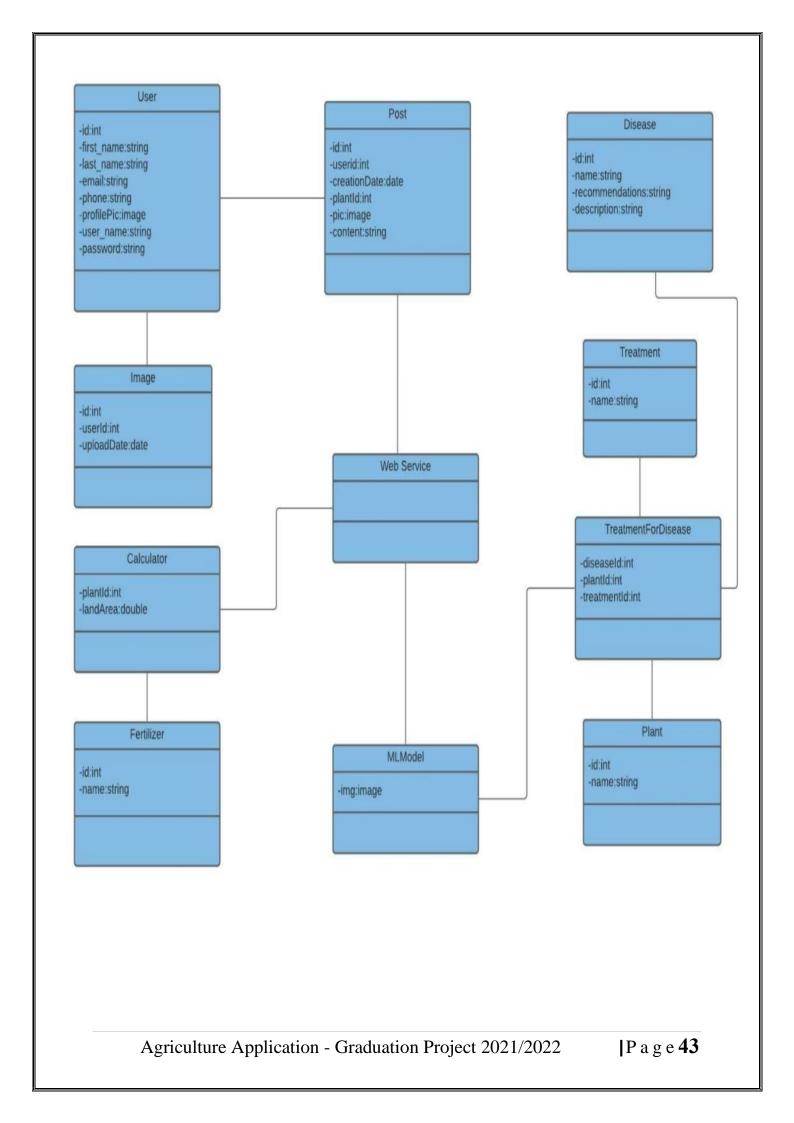


Figure 6. Class diagram





13. System Architecture

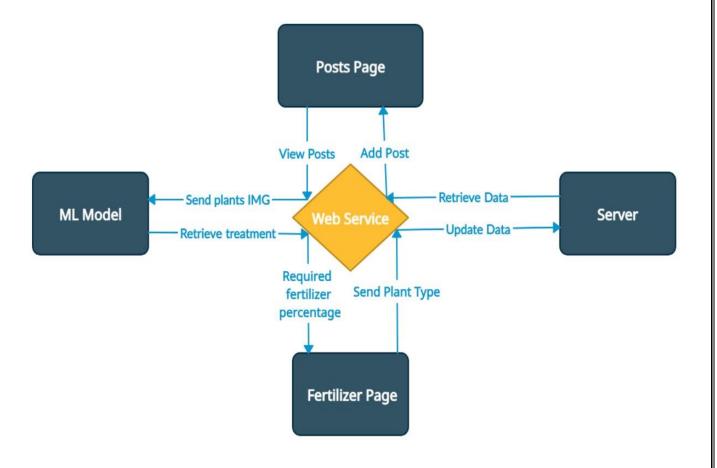


Figure 8. System architecture

- The main connection point is the Web Service (API) that connects all the system components.
- The user can make a post or can view the posts by sending a request to the API to make/view posts from/to the database server.
- User can upload the plant IMG to the API and the ML model will retrieve the treatment and the plant pest type.
- User can send the plant type and the cultivated area to the API, and it will retrieve the required fertilizer percentage.

| 14. User Interface Mockup | |
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| 15. | ER diagram | |
|-----|--|----------------|
| | Database Scheme | |
| 10. | Database Scheme | |
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