

FERTILIZERS RECOMMENDATION SYSTEM FOR DISEASE PREDICTION

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Abstract

Agriculture is one among the most important sector of Indian Economy. More than 50% of population depends on agriculture as their source of income. Growing crops over thousands of years without caring about replenishing has led to depletion and exhaustion of soil nutrients resulting in their low productivity. To improve the production, chemical fertilizers were added. But excessive use of these not only makes the plants dependent on artificial fertilizers but also erodes the natural quality of land.

It is therefore important to make sure that only the sufficient number of fertilizers is added so that yield can be increased and at the same time, the natural quality of soil remain intact. For this, it is essential to know the soil nutrient levels. Our project aims at finding the soil nutrient richness and predict the fertility of a given soil sample in real time. Based on the result obtained, the system will also be giving recommendation on the type of fertilizer and in what quantity It is to be used in the soil sample so that the yield can be maximized. Detection and recognition of plant diseases using machine learning are very efficient in providing symptoms of identifying diseases at its earliest.

1.INTRODUCTION

India is an agricultural country and depends on agricultural products for their wellbeing. It is agriculture that promotes the economic growth and the development of our country. But recently many problems have been faced by the farmers due to certain natural calamities. Apart from these major calamities, they were also in lack of sufficient knowledge about the nutrients present in their soil. Not only the soil type, the climatic condition and the usage of fertilizer also play a major role. Certain varieties of crops can be cultivated based on the climatic condition in their locality and accordingly the fertilizers can be preferred. In certain situation the usage of fertilizers also affects the cultivation. Under a climatic condition the cultivation of a right crop, usage of right fertilizer to the soil gives better yield. Mostly, fertilizers were recommended based on the nutrients present in the soil. On using chemical fertilizer the quality or the nutrients present in the soil was degraded, that promotes a decrease in the nutritive value of the soil. Another major factor to be considered is the disease in the crop cultivated. Identifying the disease in the plants and preferring appropriate fertilizer by the agriculturist to the farmers plays a major role. In earlier days, all these process were carried out manually. But with the advancement of technology the entire system was digitalized. But even then there exist various problems that need careful attention.

1.1 PROJECT OVERVIEW:

Agriculture is the main aspect of country development. Many people lead their life from agriculture field, which gives fully related to agricultural products. Plant disease, especially on leaves, is one of the major factors of reductions in both quality and quantity of the food crops. In agricultural aspects, if the plant is affected by leaf disease then it reduces the growth of the agricultural level. Finding the leaf disease is an important role of agriculture preservation. After pre-processing using a median filter, segmentation is done by Guided Active Contour method and finally, the leaf disease is identified by using Support Vector Machine. The disease-based similarity measure is used for fertilizer recommendation.

1.2 PURPOSE:

The purpose of analyzing later-Plant leaf disease identification is especially needed to predict both the quality and quantity of the First segmentation step primarily based on a mild polygonal leaf model is first achieved and later used to guide the evolution of an energetic contour. Combining global shape descriptors given by the polygonal model with local curvature. The purpose of image preprocessing is improving image statistics so that undesired distortions are suppressed and image capabilities which are probably relevant for similar processing are emphasize.

2.LITERATURE SURVEY

2.1 EXISTING PROBLEM:

[1]The author proposed a method which helps us to find the crop affected by disease and recommend a fertilizer for the crop based on the severity level. The fertilizers may be organic or inorganic. The algorithm that they used to train the model is Support Vector machine. Leaves are affected by bacteria, fungi, virus, and other insects. Support Vector Machine (SVM) algorithm classifies the leaf image as normal or affected. Vectors are constructed based on leaf features such as color, shape, textures. Then hyperplane constructed with conditions to categorize the preprocessed leaves and also implement multiclass classifier, to predict diseases in leaf image.

Advantages: The proposed SVM technique gives a better result. For the same set of images, F-Measure for CNN is 0.7and 0.8 for SVM, the accuracy of identification of leaf disease of CNN is 0.6 and SVM is 0.8.

Disadvantages: The accuracy of this algorithm can further be improved and does not provide accurate result for all diseased crops.

[2] The main objective of this paper is image analysis & classification techniques for detection of leaf diseases and classification. The leaf image is firstly preprocessed and then does the further work. K-Means Clustering used for image segmentation and then system extract the GLCM features from disease detected images. The disease classification done through the SVM classifier.

Algorithm used: Gray-Level Co-Occurrence Matrix (GLCM) features, SVM, K-Means Clustering .

Advantages: The system detects the diseases on citrus leaves with 90% accuracy.

Disadvantages: System only able to detect the disease from citrus leaves.

[3] The proposed method uses SVM to classify tree leaves, identify the disease and suggest the fertilizer. The proposed method is compared with the existing CNN based leaf disease prediction. The proposed SVM technique gives a better result when compared to existing CNN. For the same set of images, F-Measure for CNN is 0.7 and 0.8 for SVM, the accuracy of identification of leaf disease of CNN is 0.6 and SVM is 0.8.

Advantages: The prediction and diagnosing of leaf diseases are depending on the segmentation such as segmenting the healthy tissues from diseased tissues of leaves.

Disadvantages: This further research is implementing the proposed algorithm with the existing public datasets. Also, various segmentation algorithms can be implemented to improve accuracy. The proposed algorithm can be modified further to identify the disease that affects the various plant organs such as stems and fruits.

2.2 REFERENCES :

[1] Luca Bencini, Davide Di Palma, Giovanni Collodi, G. Manes and Antonio Manes, "Agricultural monitoring based on wireless sensor network technology: Real long life deployments for physiology and pathogens control.". Third International Conference on Sensor Technologies and Applications. IEEE, 2009.

[2] International Journal of Computer Science and Informatics. Jay Gholap, Anurag Ingole, Jayesh Gohil, Shailesh Gargade and Vahida Attar, "Soil Data Analysis Using Classification Techniques and Soil Attribute Prediction", IJCSI, Vol. 9, Issue 3, No 3, ISSN: 1694-0814, May 2012.

[3] Naresh, Y. G., and H. S. Nagendraswamy, "Classification of medicinal plants: an approach using modified LBP with symbolic representation", Neurocomputing 173, pp: 1789-1797, 2016.

2.3 PROBLEM STATEMENT DEFINITION :

Farmers' conventional methods of agricultural cultivation are ineffective. It does not make proper use of all available resources. Farmers are unable to detect crop diseases due to a lack of knowledge and old practices, which often result in soil nutrient deterioration and exhaustion. As a result, crop failure occurs. Growing only certain crops depletes the soil, and if the crops are harmed by illnesses, farmers are uninformed of how to recover such crops. Food needs cannot be met until and unless efficient resource management and use is implemented.



Problem Statement (PS)	I am (Customer)	I'm trying to	But	Because	Which makes me feel
PS 1	A member belongs to farmer family who is facing many struggles in cultivating the crops	Analyze their problem and identify solution for the problem. I discussed with my family. They told that the cultivation was smooth for past 10 years	In the recent years, their profit and cultivation was decreased	Crops were affected by disease due to drastic climate change. they found it difficult for finding fertilizer for the growth of crops	Frustrated so, I decided to give a solution by recommending the fertilizer for the crops which is affected by disease .This saves their cost, increase yield and productivity
PS 2	Fertilizer shop owner who provide fertilizer for the affected (diseased) crops	Give the best fertilizer for the crops which make the crop to grow in better way. One day, a farmer approached me and asked fertilizer for the diseased crop. I couldn't find the disease of the crop immediately. Later I gave a fertilizer for the crop	I came to know that the fertilizer which I provided does not reduce the disease in the plant	The fertilizer which I provided is inappropriate	That if there is any system or application which recommended accurate fertilizer for the diseased crop that increase the crop yield and save time for searching fertilizer

3.IDEATION & PROPOSED SOLUTION

3.1 EMPATHY MAP CANVAS:

A fast and exact approach to identify plant infections appears so critical for the good thing about trade and biology to agriculture. Disease control procedures can be a waste of money and time if the disease isn't properly identified and it can lead to more plant loss. Our project proposes a deep on the model will achieve its goal by categorising photos of leaves into unhealthy categories based on defect patterns.



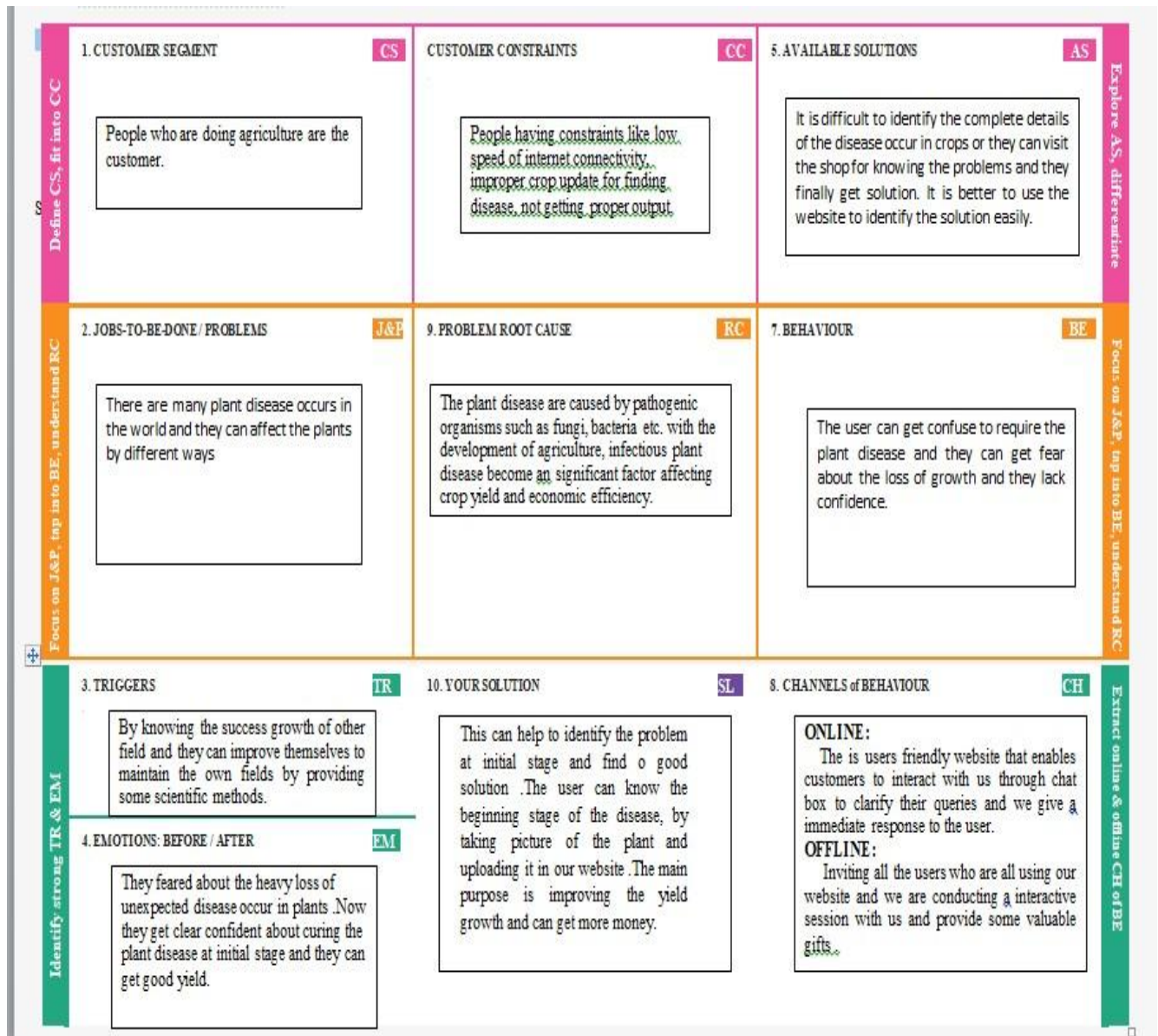
3.2 IDEATION & BRAINSTORMING

[illegible]

3.3 PROPOSED SOLUTION:

S.NO	Parameter	Description
1.	Problem Statement (Problem to be solved)	Agriculture is an extremely risky industry and our farmers are at the forefront of the industry. Farmer's face several problems which include crops get affected by diseases, the soil is not being nutritious enough for the plant to grow, etc. More than 15% of the crops get wasted in India due to diseases and hence it has become one of the major concerns to be resolved.
2.	Idea / Solution description	The solution for problem is by using Deep learning technique, which recommended the fertilizer for the affected crops. The fertilizer recommendation model provide the best recommendation for the crop based on their nutrition value, the most important issue is plant gets easily affected by diseases. We also create a website for using this application as this application is used anywhere by anyone. We have tried many different algorithms for finding better accuracy to recommend right fertilizer for affected crop.
3.	Novelty / Uniqueness	The combination of increasing global Smartphone penetration in the rural areas and recent advances in computer vision made possible by deep learning has paved the way for web assisted disease diagnosis also we have created an open source easy to use web application to address the issues which help to improve crop production.
4.	Social Impact / Customer Satisfaction	The farmer who uses this application has ability to minimize the manpower needed for recommendation. Using this model, the agriculture domain will have a great impact on the environment and there is an increase in the productivity and yield. The buyer and seller will also be more beneficial by attaining more profit
5.	Business Model (Revenue Model)	The usages of this application for predicting the fertilizers, analyzing the disease in a touch make the farmers easy with minimal subscription. This action adds a lot of value to the company and business in society
6.	Scalability of the Solution	The machine learning model which is trained using CNN have the capacity to scale when needed and operated with same speed. High model accuracy will improve for more testing and training data.

3.4.PROBLEM SOLUTION FIT:



4.REQUIREMENT ANALYSIS

4.1 FUNCTIONAL REQUIREMENT :

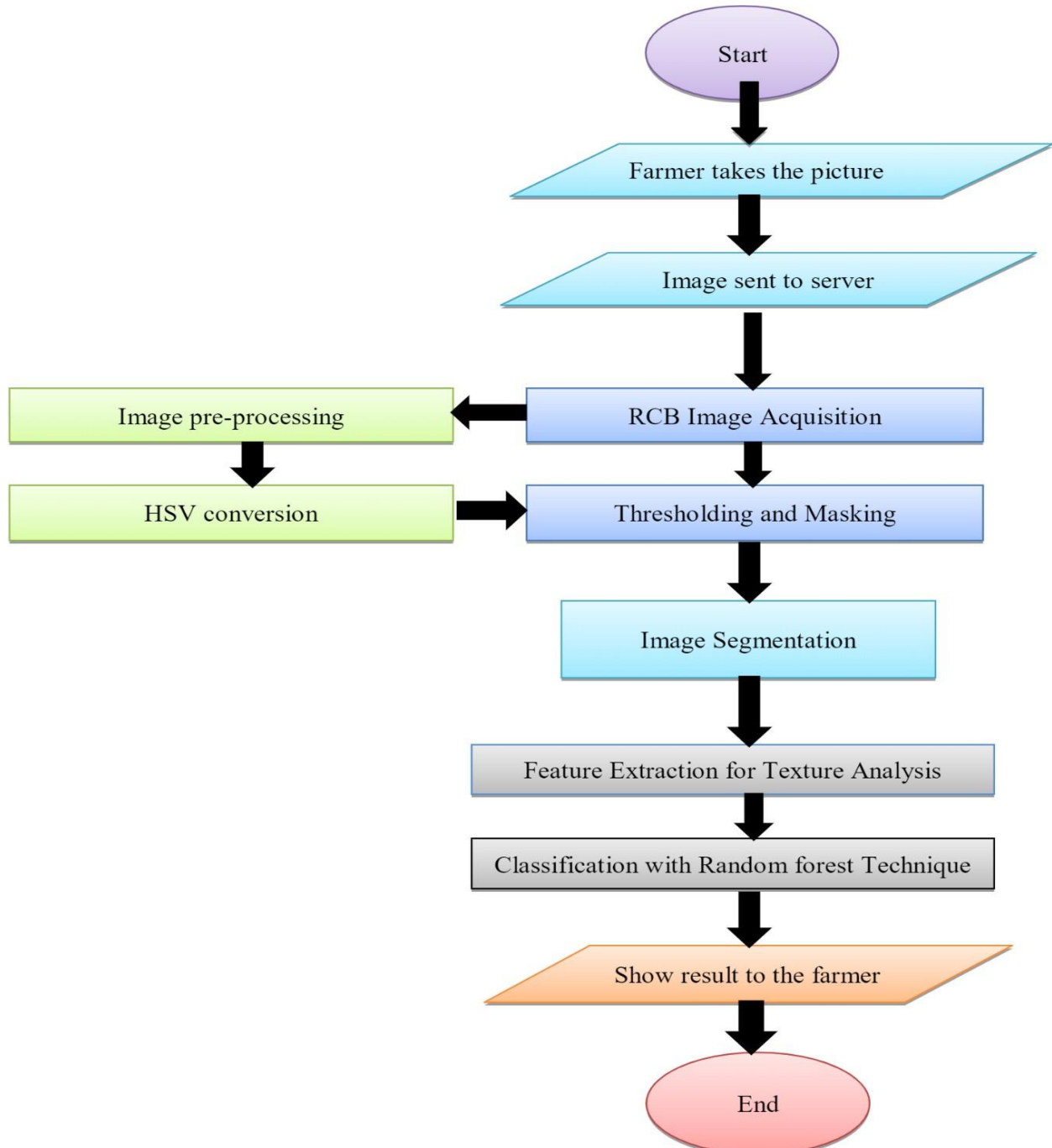
FR NO	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR – 1	User Registration	Registration through Form Registration through Gmail Registration through LinkedIn
FR - 2	User Confirmation	Confirmation via Email Confirmation via OTP
FR - 3	Upload Images for Prediction	User need to upload images for the prediction of disease in the image
FR - 4	User Requirement	The user simply give a inputs disease leaf image. The software will instantly generate and accurate disease of leaf to the farmer and also provide the suitable fertilizer to the disease leaf.

4.2 NON FUNCTIONAL REQUIREMENT:

FR No.	Non-Functional Requirement	Description
NFR-1	Usability	User can easily understand the application. They satisfied with the result.
NFR-2	Security	With the help of the user name and password it provides more security and the data are private
NFR-3	Reliability	This website must perform without failure in 98 percentage of use case
NFR-4	Performance	This application supporting 2,000 users per hour must provide 4 seconds or less response time in a desktop browse
NFR-5	Availability	User can access any time
NFR-6	Scalability	The website must be scalable enough to support 5,000 visits at the same time while maintaining optimal performance

5. PROJECT DESIGN

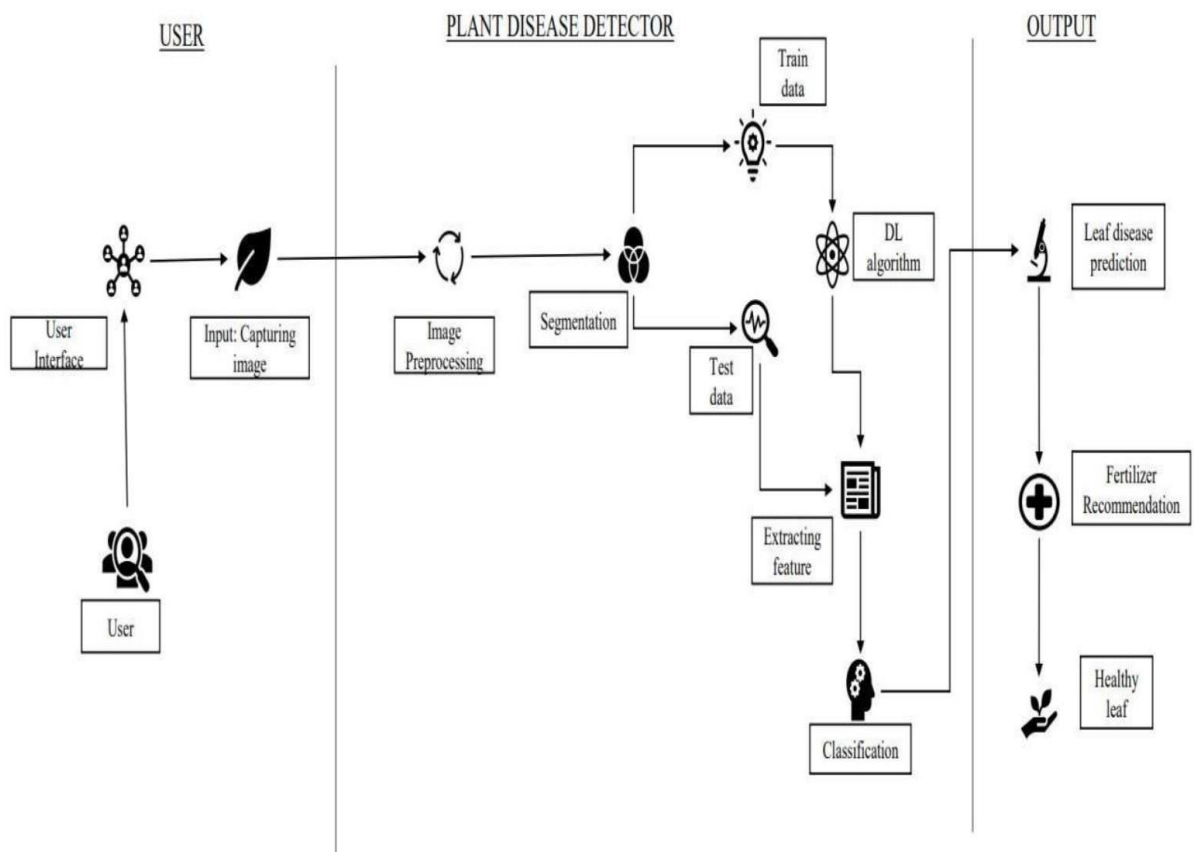
5.1 DATA FLOW DIAGRAM :



5.2 SOLUTION & TECHNICAL ARCHITECTURE:

Solution architecture is a complex process-with many sub-process-that bridges the gapbetween business problems and technology solutions. Its goals are to:

- Find the best tech solution to solve existing business problems.
- Describe the structure, characteristics, behavior, and other aspects of the software toproject stakeholders.
- Define features, development phases, and solution requirements.
- Provide specifications according to which the solution is defined, managed anddelivered.



5.3 USER STORIES :

USER TYPE	FUNCTIONAL REQUIREMENT	USER STORY NUMBER	USER STORY / TASK	ACCEPTANCE CRITERIA	PRIORITY	RELEASE
Customer	Registration	User -1	I can enter into the website and click on the register form and fill the details given in that form	I can get my profile after registering	High	High Speed
Farmer	Login	User-2	As a user ,I can login by using email ,password and confirming password	I can receive confirmation email & click confirm	High	High Speed
Agricultural workers	Sign in	User -3	As a user ,I can sign in by filling some details like name, address , phone number in that chat	After sign in I have a new account	Low	Take a minute

6.PROJECT PLANNING & SCHEDULING

6.1 SPRINT PLANNING AND ESTIMATION:

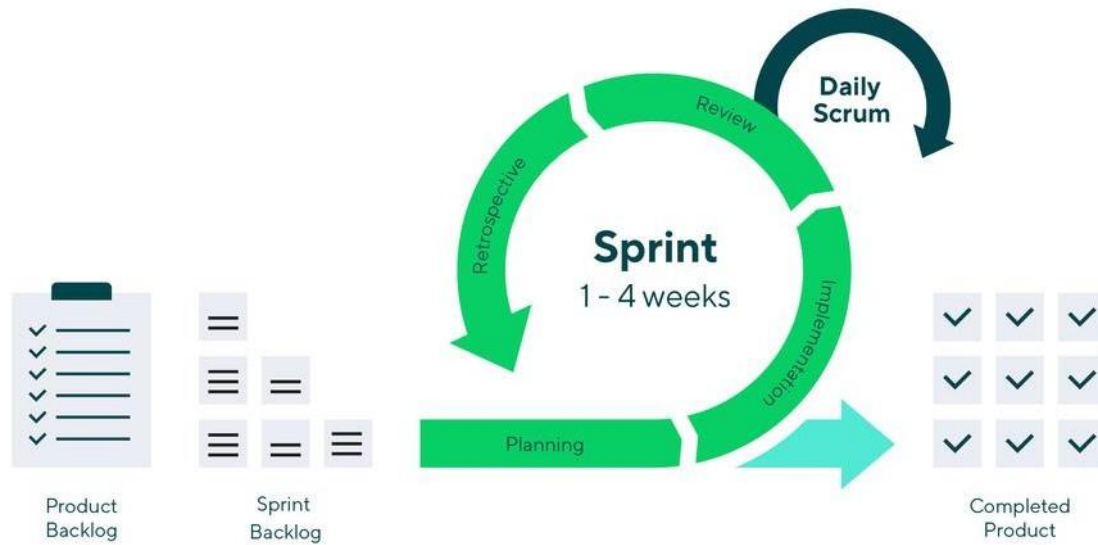
Sprint Planning:

The performance of Artificial Intelligence (AI) models is being improved and increased in modern technology.

- Based Crop Yield Disease Prediction System would assist farmers in protecting their crops from a variety of diseases by identifying them during the process of taking an image at the plant and providing the afflicted disease's name to a machine learning algorithm.
- The best answer for the farmer will be provided in this project milestone, and he or she may find it on their own by using a web application with a completely user-friendly and straightforward user interface.
- Additionally, we intend to add a useful Module that is a fertilizer prescription for a certain disease to the process. It can propose both artificial and natural fertilizer in a similar way.

Estimation:

1. Planning is a crucial role in project management because it allows team Members to schedule their time on the project.
2. This activity demonstrates how the team members assigned and completed various tasks!
3. In Project we can Split into the Four Step of Phases are,
 - Phase 1: Information Collection and Requirement Analysis
 - Phase 2: Project Planning and Developing Modules
 - Phase 3: Implementing the High Accuracy Deep Learning Algorithm to Perform
 - Phase 4: Deploying the Model on Cloud and Testing the Model and UI Performance.



6.2 SPRINT DELIVERY SCHEDULE

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points (Total)	Priority	Team Members
Sprint-1	Collection of Dataset		Collecting all the required dataset that are used to train and test the model	4	High	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
	Model Creation and Training (Fruits)		Create a model which can classify diseased fruit plants from healthy plants from the images. I also need to test the model and deploy it on IBM Cloud.	4	High	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
	Model Creation and Training (Vegetables)		Create a model which can classify diseased vegetable plants from healthy plants from the given images and also testing the model by deploying it in IBM Cloud.	4	High	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points (Total)	Priority	Team Members
Sprint-2	Model Training and testing in IBM Cloud		Create a model which can classify diseased vegetable plants from given images and train on IBM Cloud	6	High	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
	Registration	USN-1	As a user, I can register by entering my email, password, and confirming my password or via OAuth API	3	Medium	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
	Upload page	USN-2	As a user, I will be redirected to a page where I can select based on my requirement whether to upload my pictures of crops for disease prediction or entering my soil details for crop recommendation or fertilizer recommendation.	4	High	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
	Suggestion results	USN-3	As a user, I can view the results and then obtain the suggestions provided by the ML model	4	High	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
	Base Flask App		A base Flask web app must be created as an interface for the ML model to interact.	2	High	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
Sprint-3	Login	USN-4	As a user/admin/shopkeeper, I can log into the application by entering email & password	2	High	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
	User Dashboard	USN-5	As a user, I can view the previous results and history which saves the user's time.	3	Medium	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V
	Integration		Integrate Flask, CNN model with Cloud and Database.	5	Medium	Akalya.S, Devika.P, Dhivyadharshini.S, Manju.V

	Containerization		Containerize Flask app using Docker	5	Low	Akalya.S, Devika.P, Dhivyadharshini.S , Manju.V
Sprint -4	Testing and Documentation		Finally the project is tested and further improvements is made based on user feedback. Documentation is also made in order to make better user experience.	4	Medium	Akalya.S, Devika.P, Dhivyadharshini.S , Manju.V
	Dashboard	USN-6	As a shopkeeper, I can enter fertilizer products and then update the details if any	4	Low	Akalya.S, Devika.P, Dhivyadharshini.S , Manju.V
	Containerization		Create and deploy Helm charts using Docker Image made before.	2	Low	Akalya.S, Devika.P, Dhivyadharshini.S , Manju.V

Project Tracker, Velocity & Burndown Chart:

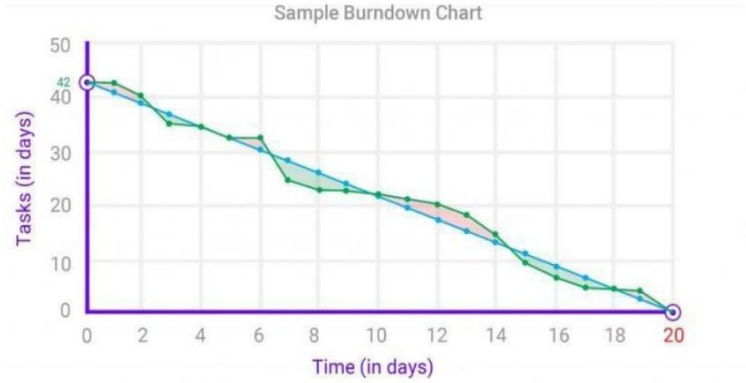
Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	10	6 Days	24 Oct 2022	29 Oct 2022	10	30 Oct 2022
Sprint-2	15	6 Days	31 Oct 2022	05 Nov 2022	15	07 Nov 2022
Sprint-3	15	6 Days	07 Nov 2022	12 Nov 2022	15	13 Nov 2022
Sprint-4	12	6 Days	14 Nov 2022	19 Nov 2022	10	20 Nov 2022

VELOCITY:

Imagine we have a 10 days sprint duration, and the velocity of the team is 20. Let's calculate the team's average velocity per iteration unit.

$$AV = \frac{\text{sprint duration}}{\text{velocity}} = \frac{20}{10} = 2$$

BURNDOWN CHART:



ROADMAP:

Jira Software

Your work

Projects

Filters

Dashboards

People

Apps

Create

Q Search

?

AS

Fertilizer Recommend...

Software project

PLANNING

Roadmap

Backlog

Board

Reports

DEVELOPMENT

Code

Project pages

IBM-Project-15473-1...

You're in a team-managed project

Learn more

Projects / Fertilizer Recommendation System for Disease Prediction

FRSFDP board

Q

AS

GROUP BY

Subtask

TO DO

IN PROGRESS

DONE ✓

+

Everything else no issues

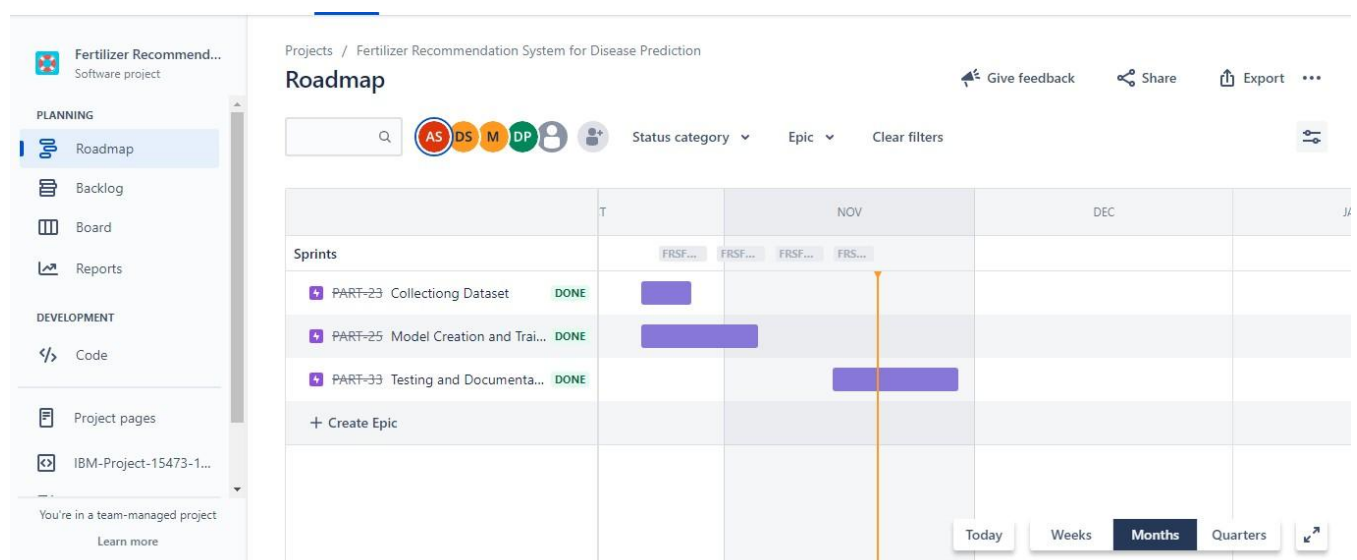
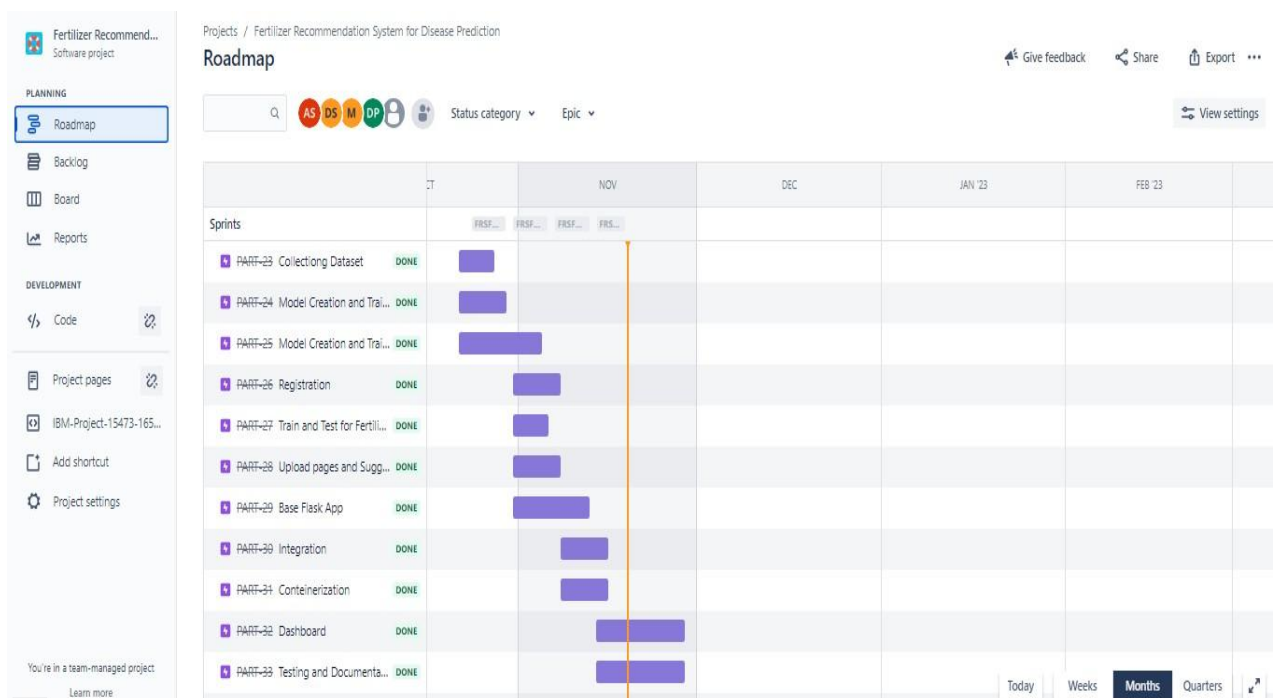
You haven't started a sprint

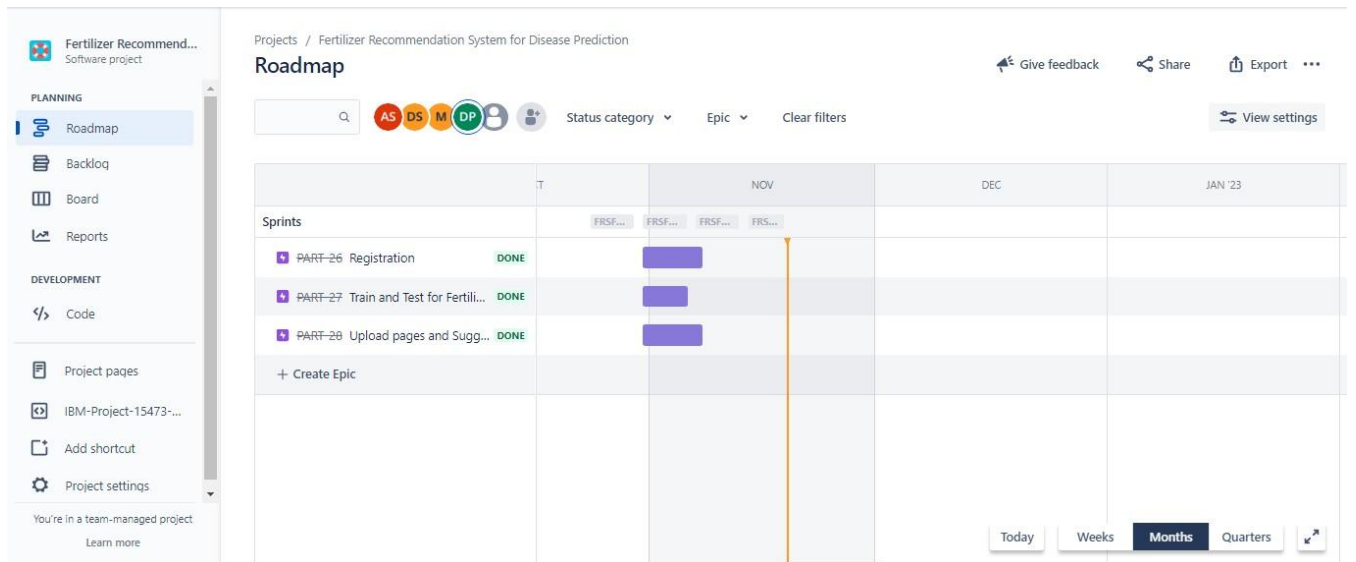
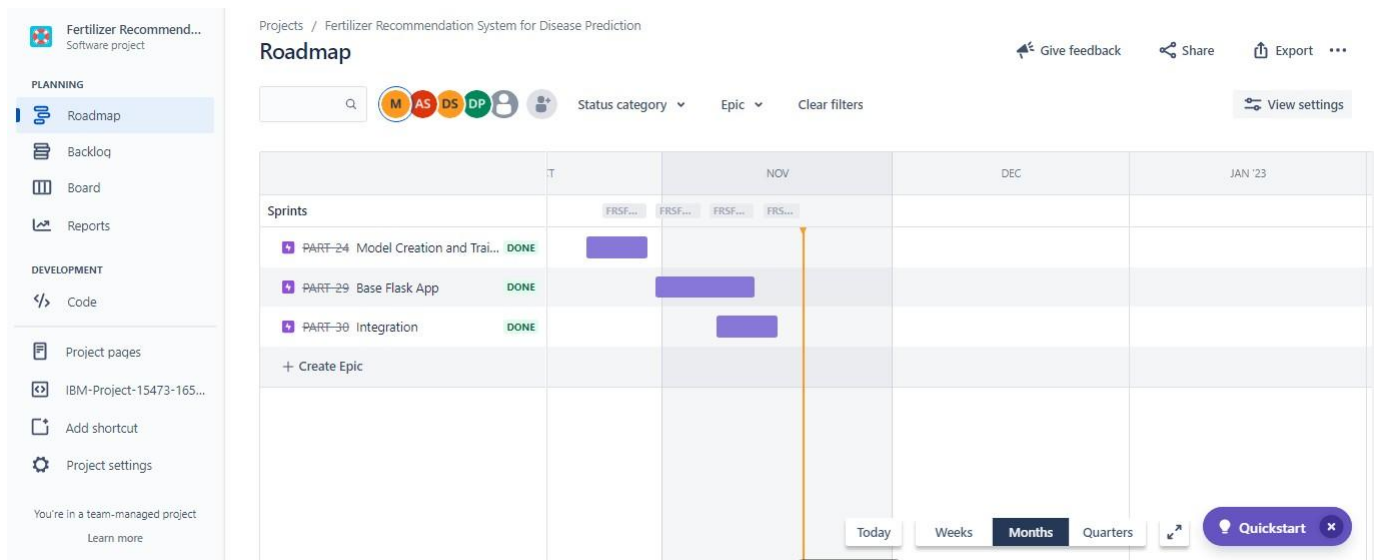
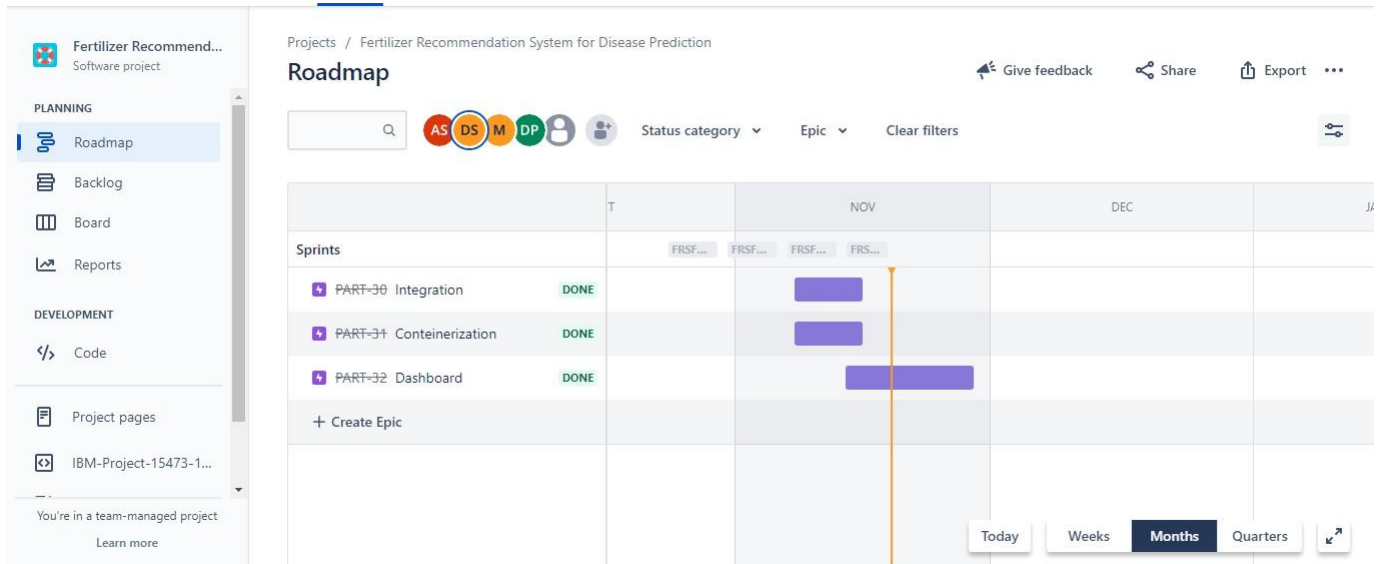
You can't do anything on your board because you haven't started a sprint yet. Go to the backlog to plan and start a sprint.

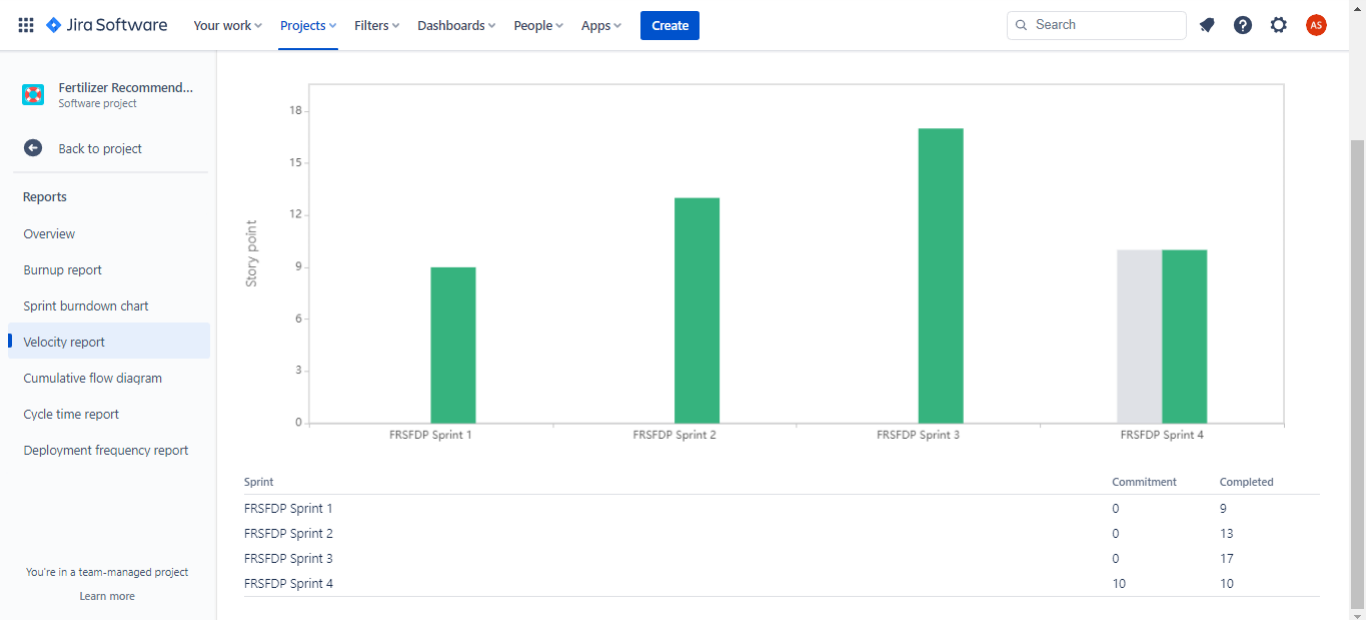
6.3 REPORTS FROM JIRA

Jira Software is part of a family of products designed to help teams of all types manage work. Originally, Jira was designed as a bug and issue tracker. But today, Jira has evolved into a powerful work management tool for all kinds of use cases, from requirements and test case management to agile software development.

Jira is one of the best open-source tools for planning and tracking in Agile methodology. Development teams use Jira for tracking bugs and projects, managing Scrums, and visualizing workflows with Kanban boards. Workflows in Jira make it easy to plan, track, release, and report on software.







7.CODING AND SOLUTIONING



The block diagram of the entire project is shown in figure. First step is the image dataset collection followed by image preprocessing. The third step is the training of image datasets with initializing different hyper parameters. Then build the model and save the model file with .pth format. The final stage is the testing of existing or new datasets using the trained model.

HARDWARE/SOFTWARE DESIGNING

The software used for training and testing the dataset is Python. The Jupyter notebook (Notebook of IBM cloud also) is used for python programming. The neural network used for training and testing the model is Convolutional Neural Network(CNN). The CNN has following layers:

- Convolutional layer (32x32 kernel (3x3))
- Max-pool layer (kernel(2x2))
- Flatten layer
- Dense layer (different layers with different size)
- Final output dense layer(size 6x1 for fruit dataset and 9x1 for Vegetable dataset)

In the preprocessing step, images are normalized to 1 and then resized to 128x128. The images are arranged in different batch sizes. Then train set and test set are formed from the collected datasets. In order to do the above steps in Python, the following Python libraries must be imported before starting the process:

- Numpy
- TensorFlow
- Keras
- Matplotlib (optional for data visualization)

The following activation functions used in the CNN training:

- RELU at the end of convolution layer and Max Pool layer
- Softmax at the end of output dense layer
- For testing the dataset argmax is used, its an optional

DATASETLINK:

Dataset: <https://www.kaggle.com/datasets/vipooool/new-plant-diseases-dataset>

IMAGE PREPROSSESSING

Image preprocessing is the steps taken to format images before they are used by model training and inference. This includes, but is not limited to, resizing, orienting, and color corrections. Image preprocessing may also decrease model training time and increase model inference speed. If input images are particularly large, reducing the size of these images will dramatically improve model training time without significantly reducing model performance. For example, the standard size of images on iPhone 11 is 3024×4032 . The machine learning model Apple uses to create masks and apply Portrait Mode performs on images half this size before its output is rescaled back to full size.

Preprocessing is an essential step to clean image data before it is ready to be used in a computer vision model. There are both technical and performance reasons why preprocessing is essential. Fully connected layers in convolution neural networks, a common architecture in computer vision, require that all images are the same sized arrays. If your images are not in the same size, your model may not perform as expected. If you are building a model in code using a library like Tensor flow, you are likely to encounter an error if your image is not the same size.

FEATURE 1

IMAGE PREPOSSESSING CODING FOR FRUIT

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
train_datagen = ImageDataGenerator(rescale = 1./255,  
  
    shear_range = 0.2,  
  
    zoom_range = 0.2,  
  
    horizontal_flip = True)
```

```
test_datagen = ImageDataGenerator(rescale = 1./255)
```

```
training_set =  
train_datagen.flow_from_directory('/content/drive/MyDrive/plant_dataset/fruit_  
dataset/train',target_size=(224,224),batch_size=32,class_mode='categorical')
```

```
test_set =  
train_datagen.flow_from_directory('/content/drive/MyDrive/plant_dataset/fruit_  
dataset/test',target_size=(224,224),batch_size=32,class_mode='categorical')
```

IMAGE PREPROSSCESSING CODING FOR VEGETABLE

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
train_datagen = ImageDataGenerator(rescale = 1./255,  
  
    shear_range = 0.2,  
  
    zoom_range = 0.2,  
  
    horizontal_flip = True)
```

```
test_datagen = ImageDataGenerator(rescale = 1./255)
```

```
training_set =  
train_datagen.flow_from_directory('/content/drive/MyDrive/plant_dataset/fruit_  
dataset/train',target_size=(224,224),batch_size=32,class_mode='categorical')
```

```
test_set =  
train_datagen.flow_from_directory('/content/drive/MyDrive/plant_dataset/fruit_  
dataset/test',target_size=(224,224),batch_size=32,class_mode='categorical')
```

MODEL BULIDING:

```
ls  
  
cd /content/drive/MyDrive/full_dataset  
  
ls
```

Mounting google drive

```
from google.colab import drive  
  
drive.mount('/content/drive')
```

Installing torch

```
!pip install torchsummary
```

Import necessary libraries

```
import os  
  
import numpy as np  
  
import pandas as pd  
  
import torch  
  
import matplotlib.pyplot as plt  
  
import torch.nn as nn  
  
from torch.utils.data import DataLoader  
  
from PIL import Image  
  
import torch.nn.functional as F  
  
from torchvision import datasets, transforms, models  
  
import torchvision.models as models  
  
from torchvision.utils import make_grid  
  
from torchvision.datasets import ImageFolder
```

```
from torchsummary import summary

from matplotlib.ticker import FormatStrFormatter

%matplotlib inline
```

Unzip the file

```
!unzip archive.zip
```

Exploring the data

```
data_dir = "/content/drive/MyDrive/full_dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)"

train_dir = data_dir + "/train"

valid_dir = data_dir + "/valid"

diseases_plant = os.listdir(train_dir)

diseases_plant

plants = []

No_Of_Diseases = 0

for plant in diseases_plant:

    if plant.split('___')[0] not in plants:

        plants.append(plant.split('___')[0])

    if plant.split('___')[0] != 'healthy':

        No_Of_Diseases += 1

print(plants)

len(plants)

format(No_Of_Diseases)

nums = { }

for disease in diseases_plant:

    nums[disease] = len(os.listdir(train_dir + '/' + disease))
```

```
img_per_class = pd.DataFrame(nums.values(), index=nums.keys(),
columns=["number_of_images"])
```

```
img_per_class
```

```
X_train = 0
```

```
for value in nums.values():
```

```
    X_train += value
```

```
print(X_train)
```

datasets for validation and training

```
train = ImageFolder(train_dir, transform=transforms.ToTensor())
```

```
valid = ImageFolder(valid_dir, transform=transforms.ToTensor())
```

```
img, label = train[0]
```

```
print(img.shape, label)
```

```
len(train.classes)
```

for checking some images from training dataset

```
def show_image(image, label):
```

```
    print("Label :" + train.classes[label] + "(" + str(label) + ")")
```

```
    plt.imshow(image.permute(1, 2, 0))
```

```
show_image(*train[101])
```

```
show_image(*train[30500])
```

```
show_image(*train[50000])
```

```
random_seed = 7
```

```
torch.manual_seed(random_seed)
```

setting the batch size

```
batch_size = 32
```

DataLoaders for training and validation

```
train_dl = DataLoader(train, batch_size, shuffle=True, num_workers=2,
pin_memory=True)
```



```
valid_dl = DataLoader(valid, batch_size, num_workers=2, pin_memory=True)
```

```
def show_batch(data):
```

```
    for images, labels in data:
```

```
        fig, ax = plt.subplots(figsize=(30, 30))
```

```
        ax.set_xticks([]); ax.set_yticks([])
```

```
        ax.imshow(make_grid(images, nrow=8).permute(1, 2, 0))
```

```
        break
```

Images for first batch of training

```
show_batch(train_dl)
```

for moving data into GPU (if available)

```
def get_default_device():
```

```
    """Pick GPU if available, else CPU"""
```

```
    if torch.cuda.is_available:
```

```
        return torch.device("cuda")
```

```
    else:
```

```
        return torch.device("cpu")
```

```
#for moving data to device (CPU or GPU)
```

```
def to_device(data, device):
```

```
    """Move tensor(s) to chosen device"""
```

```
    if isinstance(data, (list,tuple)):
```

```
        return [to_device(x, device) for x in data]
```

```
    return data.to(device, non_blocking=True)
```

```
# for loading in the device (GPU if available else CPU)
```

```
class DeviceDataLoader():
```

```

        """Wrap a dataloader to move data to a device"""

    def __init__(self, dl, device):

        self.dl = dl

        self.device = device


    def __iter__(self):

        """Yield a batch of data after moving it to device"""

        for b in self.dl:

            yield to_device(b, self.device)


    def __len__(self):

        """Number of batches"""

        return len(self.dl)

device = get_default_device()

device

# Moving data into GPU

train_dl = DeviceDataLoader(train_dl, device)

valid_dl = DeviceDataLoader(valid_dl, device)

class SimpleResidualBlock(nn.Module):

    def __init__(self):

        super().__init__()

        self.conv1 = nn.Conv2d(in_channels=3, out_channels=3, kernel_size=3,
stride=1, padding=1)

        self.relu1 = nn.ReLU()

        self.conv2 = nn.Conv2d(in_channels=3, out_channels=3, kernel_size=3,
stride=1, padding=1)

        self.relu2 = nn.ReLU()

```

```
def forward(self, x):  
    out = self.conv1(x)  
    out = self.relu1(out)  
    out = self.conv2(out)  
    return self.relu2(out) + x
```

for calculating the accuracy

```
def accuracy(outputs, labels):  
    _, preds = torch.max(outputs, dim=1)  
    return torch.tensor(torch.sum(preds == labels).item() / len(preds))
```

base class for the model

```
class ImageClassificationBase(nn.Module):  
    def training_step(self, batch):  
        images, labels = batch  
        out = self(images)          # Generate predictions  
        loss = F.cross_entropy(out, labels) # Calculate loss  
        return loss  
  
    def validation_step(self, batch):  
        images, labels = batch  
        out = self(images)          # Generate prediction  
        loss = F.cross_entropy(out, labels) # Calculate loss  
        acc = accuracy(out, labels)      # Calculate accuracy  
        return {"val_loss": loss.detach(), "val_accuracy": acc}  
  
    def validation_epoch_end(self, outputs):  
        batch_losses = [x["val_loss"] for x in outputs]  
        batch_accuracy = [x["val_accuracy"] for x in outputs]
```

```

epoch_loss = torch.stack(batch_losses).mean()    # Combine loss

epoch_accuracy = torch.stack(batch_accuracy).mean()

return {"val_loss": epoch_loss, "val_accuracy": epoch_accuracy} #
Combine accuracies

def epoch_end(self, epoch, result):

    print("Epoch  [{ }], last_lr: {:.5f}, train_loss: {:.4f}, val_loss: {:.4f},
val_acc: {:.4f}".format(

        epoch,      result['lrs'][-1],      result['train_loss'],      result['val_loss'],
result['val_accuracy']))

```

Architecture for training

convolution block with BatchNormalization

```

def ConvBlock(in_channels, out_channels, pool=False):

    layers = [nn.Conv2d(in_channels, out_channels, kernel_size=3, padding=1),

nn.BatchNorm2d(out_channels),

nn.ReLU(inplace=True)]

    if pool:

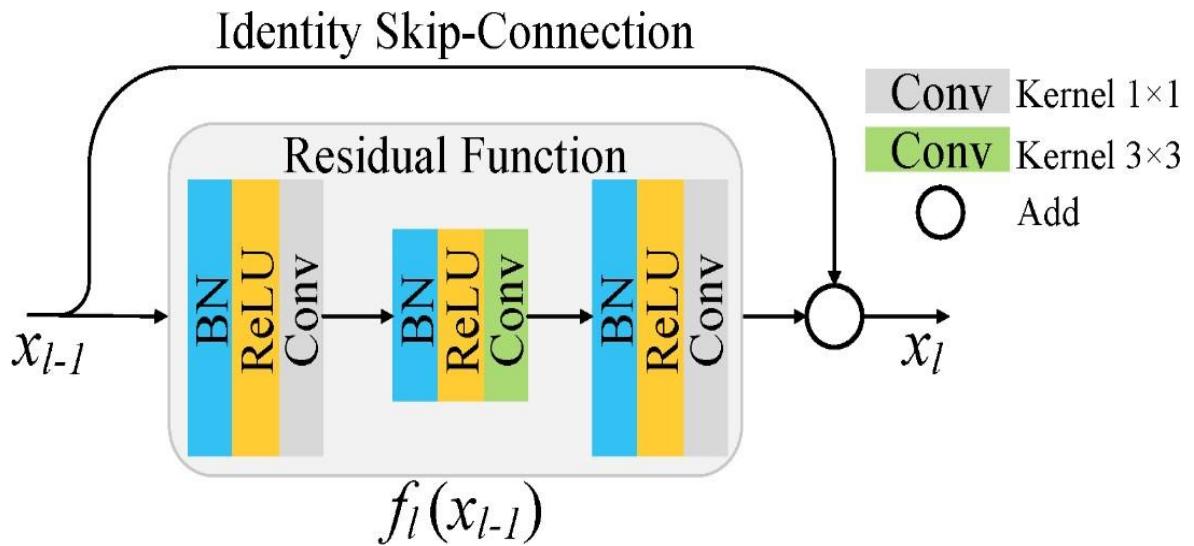
        layers.append(nn.MaxPool2d(4))

    return nn.Sequential(*layers)

```

Resnet architecture:

We are going to use ResNet, which have been one of the major breakthrough in computer vision since they were introduced in 2015. In ResNets, unlike in traditional neural networks, each layer feeds into the next layer, we use a network with residual blocks, each layer feeds into the next layer and directly into the layers about 2–3 hops away, to avoid over-fitting (a situation when validation loss stop decreasing at a point and then keeps increasing while training loss still decreases). This also helps in preventing vanishing gradient problem and allow us to train deep neural networks. Here is a simple residual block:



```
class ResNet9(ImageClassificationBase):
```

```
    def __init__(self, in_channels, num_diseases):
```

```
        super().__init__()
```

```
        self.conv1 = ConvBlock(in_channels, 64)
```

```
        self.conv2 = ConvBlock(64, 128, pool=True) # out_dim : 128 x 64 x 64
```

```
        self.res1 = nn.Sequential(ConvBlock(128, 128), ConvBlock(128, 128))
```

```
        self.conv3 = ConvBlock(128, 256, pool=True) # out_dim : 256 x 16 x 16
```

```
        self.conv4 = ConvBlock(256, 512, pool=True) # out_dim : 512 x 4 x 44
```

```
        self.res2 = nn.Sequential(ConvBlock(512, 512), ConvBlock(512, 512))
```

```
        self.classifier = nn.Sequential(nn.MaxPool2d(4),
```

```
                                         nn.Flatten(),
```

```
                                         nn.Linear(512, num_diseases))
```

```
    def forward(self, xb): # xb is the loaded batch
```

```
        out = self.conv1(xb)
```

```
        out = self.conv2(out)
```

```
out = self.res1(out) + out
```

```
out = self.conv3(out)
```

```
out = self.conv4(out)
```

```
out = self.res2(out) + out
```

```
out = self.classifier(out)
```

```
return out
```

for training

```
@torch.no_grad()
```

```
def evaluate(model, val_loader):
```

```
    model.eval()
```

```
    outputs = [model.validation_step(batch) for batch in val_loader]
```

```
    return model.validation_epoch_end(outputs)
```

```
def get_lr(optimizer):
```

```
    for param_group in optimizer.param_groups:
```

```
        return param_group['lr']
```

```
def fit_OneCycle(epochs, max_lr, model, train_loader, val_loader,  
weight_decay=0,
```

```
grad_clip=None, opt_func=torch.optim.SGD):
```

```
    torch.cuda.empty_cache()
```

```
    history = []
```

```
    optimizer = opt_func(model.parameters(), max_lr,  
weight_decay=weight_decay)
```

```
    sched = torch.optim.lr_scheduler.OneCycleLR(optimizer, max_lr,
epochs=epochs, steps_per_epoch=len(train_loader))
```

```
for epoch in range(epochs):
```

```
    # Training
```

```
    model.train()
```

```
    train_losses = []
```

```
    lrs = []
```

```
    for batch in train_loader:
```

```
        loss = model.training_step(batch)
```

```
        train_losses.append(loss)
```

```
        loss.backward()
```

```
        # gradient clipping
```

```
        if grad_clip:
```

```
            nn.utils.clip_grad_value_(model.parameters(), grad_clip)
```

```
        optimizer.step()
```

```
        optimizer.zero_grad()
```

```
    recording and updating learning rates
```

```
    lrs.append(get_lr(optimizer))
```

```
    sched.step()
```

```
    validation
```

```
    result = evaluate(model, val_loader)
```

```
    result['train_loss'] = torch.stack(train_losses).mean().item()
```

```

        result['lrs'] = lrs

        model.epoch_end(epoch, result)

        history.append(result)

    return history

%%time

history = [evaluate(model, valid_dl)]

history

epochs = 2

max_lr = 0.01

grad_clip = 0.1

weight_decay = 1e-4

opt_func = torch.optim.Adam

calculating accuracy

%%time

history += fit_OneCycle(epochs, max_lr, model, train_dl, valid_dl,

                        grad_clip=grad_clip,

                        weight_decay=1e-4,

                        opt_func=opt_func)

def plot_accuracies(history):

    accuracies = [x['val_accuracy'] for x in history]

    plt.plot(accuracies, '-x')

    plt.xlabel('epoch')

```



```

plt.ylabel('accuracy')

plt.title('Accuracy vs. No. of epochs')

def plot_lrs(history):

    lrs = np.concatenate([x.get('lrs', []) for x in history])

    plt.plot(lrs)

    plt.xlabel('Batch no.')

    plt.ylabel('Learning rate')

    plt.title('Learning Rate vs. Batch no.');
```

plot_accuracies(history)

plot_lrs(history)

Testing

```

test_dir = "/content/drive/MyDrive/full_dataset/New Plant Diseases
Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/valid"

test = ImageFolder(test_dir, transform=transforms.ToTensor())

test_images = sorted(os.listdir(test_dir)) # since images in test folder are in
alphabetical order

test_images

def predict_image(img, model):

    """Converts image to array and return the predicted class

    with highest probability"""

    # Convert to a batch of 1

    xb = to_device(img.unsqueeze(0), device)
```

```
yb = model(xb)

# Pick index with highest probability

_, preds = torch.max(yb, dim=1)

# Retrieve the class label

return train.classes[preds[0].item()]

# predicting first image

img, label = test[0]

plt.imshow(img.permute(1, 2, 0))

print('Label:', test_images[0], ', Predicted:', predict_image(img, model))

# getting all predictions (actual label vs predicted)

for i, (img, label) in enumerate(test):

    print('Label:', test[i], ', Predicted:', predict_image(img, model))
```

saving to the kaggle working directory

```
PATH = 'plant-disease-model.pth'

torch.save(model.state_dict(), PATH)
```

saving the entire model to working directory

```
PATH = '/content/drive/MyDrive/full_dataset/plant-disease-model-complete.pth'

torch.save(model, PATH)
```

FEATURE 2

APPLICATION BUILDING

```
from flask import Flask, render_template, request, Markup
```

```
import numpy as np
```

```
import pandas as pd
```

```
from utils.disease import disease_dic
```

```
from utils.fertilizer import fertilizer_dic
```

```
import requests
```

```
import config
```

```
import pickle
```

```
import io
```

```
import torch
```

```
from torchvision import transforms
```

```
from PIL import Image
```

```
from utils.model import ResNet9
```

```
import os
```

```
import requests,json
```

-----**LOADING THE TRAINED MODELS**-----

Loading plant disease classification model:

```
disease_classes = ['Apple___Apple_scab',
```

```
                   'Apple___Black_rot',
```

```
                   'Apple___Cedar_apple_rust',
```

'Apple___healthy',

'Blueberry___healthy',

'Cherry_(including_sour)___Powdery_mildew',

'Cherry_(including_sour)___healthy',

'Corn_(maize)___Cercospora_leaf_spot Gray_leaf_spot',

'Corn_(maize)___Common_rust_',

'Corn_(maize)___Northern_Leaf_Blight',

'Corn_(maize)___healthy',

'Grape___Black_rot',

'Grape___Esca_(Black_Measles)',

'Grape___Leaf_blight_(Isariopsis_Leaf_Spot)',

'Grape___healthy',

'Orange___Haunglongbing_(Citrus_greening)',

'Peach___Bacterial_spot',

'Peach___healthy',

'Pepper,_bell___Bacterial_spot',

'Pepper,_bell___healthy',

'Potato___Early_blight',

'Potato___Late_blight',

'Potato___healthy',

'Raspberry___healthy',

'Soybean___healthy',

```
'Squash___Powdery_mildew',  
'Strawberry___Leaf_scorch',  
'Strawberry___healthy',  
'Tomato___Bacterial_spot',  
'Tomato___Early_blight',  
'Tomato___Late_blight',  
'Tomato___Leaf_Mold',  
'Tomato___Septoria_leaf_spot',  
'Tomato___Spider_mites Two-spotted_spider_mite',  
'Tomato___Target_Spot',  
'Tomato___Tomato_Yellow_Leaf_Curl_Virus',  
'Tomato___Tomato_mosaic_virus',  
'Tomato___healthy']
```

```
disease_model_path = r'E:/IBM/main project/full project/app/models/plant-  
disease-model.pth'
```

```
disease_model = ResNet9(3, len(disease_classes))
```

```
disease_model.load_state_dict(torch.load(  
    disease_model_path, map_location=torch.device('cpu')))
```

```
disease_model.eval()
```

Loading crop recommendation model:

```
crop_recommendation_model_path = 'models/RandomForest.pkl'
```

```
crop_recommendation_model = pickle.load(  

```

```
open(crop_recommendation_model_path, 'rb'))
```

Custom functions for calculations:

```
def weather_fetch(city_name):
```

```
    """
```

```
    Fetch and returns the temperature and humidity of a city
```

```
    :params: city_name
```

```
    :return: temperature, humidity
```

```
    """
```

```
    api_key = "9d7cde1f6d07ec55650544be1631307e"
```

```
    base_url = "http://api.openweathermap.org/data/2.5/weather?"
```

```
    complete_url = base_url + "appid=" + api_key + "&q=" + city_name
```

```
    response = requests.get(complete_url)
```

```
    x = response.json()
```

```
    if x["cod"] != "404":
```

```
        y = x["main"]
```

```
        temperature = round((y["temp"] - 273.15), 2)
```

```
        humidity = y["humidity"]
```

```
        return temperature, humidity
```

```
    else:
```

```
        return None
```

```
def predict_image(img, model=disease_model):
```

```
    """
```

Transforms image to tensor and predicts disease label

:params: image

:return: prediction (string)

"""

transform = transforms.Compose([

transforms.Resize(256),

transforms.ToTensor(),

])

image = Image.open(io.BytesIO(img))

img_t = transform(image)

img_u = torch.unsqueeze(img_t, 0)

Get predictions from model:

yb = model(img_u)

Pick index with highest probability

_, preds = torch.max(yb, dim=1)

prediction = disease_classes[preds[0].item()]

Retrieve the class label

return prediction

FLASK APP

app = Flask(__name__)

render home page:

@ app.route('/')

```
def home():
```

```
    title = 'Heck of Harvest- Home'
```

```
    return render_template('index.html', title=title)
```

render crop recommendation form page:

```
@ app.route('/crop-recommend')
```

```
def crop_recommend():
```

```
    title = 'Heck of Harvest - Crop Recommendation'
```

```
    return render_template('crop.html', title=title)
```

render fertilizer recommendation form page

```
@ app.route('/fertilizer')
```

```
def fertilizer_recommendation():
```

```
    title = 'Heck of Harvest - Fertilizer Suggestion'
```

```
    return render_template('fertilizer.html', title=title)
```

render disease prediction input page

_____RENDER PREDICTION PAGE_____

render crop recommendation result page:

```
@ app.route('/crop-predict', methods=['POST'])
```

```
def crop_prediction():
```

```
    title = 'Heck of Harvest - Crop Recommendation'
```

```
    if request.method == 'POST':
```



```

N = int(request.form['nitrogen'])

P = int(request.form['phosphorous'])

K = int(request.form['pottasium'])

ph = float(request.form['ph'])

rainfall = float(request.form['rainfall'])

# state = request.form.get("stt")

city = request.form.get("city")

if weather_fetch(city) != None:

    temperature, humidity = weather_fetch(city)

    data = np.array([[N, P, K, temperature, humidity, ph, rainfall]])

    my_prediction = crop_recommendation_model.predict(data)

    final_prediction = my_prediction[0]

    return render_template('crop-result.html', prediction=final_prediction,
title=title)

else:

    return render_template('try-again.html', title=title)

```

render fertilizer recommendation result page:

```

@ app.route('/fertilizer-predict', methods=['POST'])

def fert_recommend():

    title = 'Heck of Harvest - Fertilizer Suggestion'

    crop_name = str(request.form['cropname'])

    N = int(request.form['nitrogen'])

```

```
P = int(request.form['phosphorous'])

K = int(request.form['pottasium'])

# ph = float(request.form['ph'])

df = pd.read_csv('data/fertilizer.csv')

nr = df[df['Crop'] == crop_name]['N'].iloc[0]

pr = df[df['Crop'] == crop_name]['P'].iloc[0]

kr = df[df['Crop'] == crop_name]['K'].iloc[0]

n = nr - N

p = pr - P

k = kr - K

temp = {abs(n): "N", abs(p): "P", abs(k): "K"}

max_value = temp[max(temp.keys())]

if max_value == "N":

    if n < 0:

        key = 'NHigh'

    else:

        key = "Nlow"

elif max_value == "P":

    if p < 0:

        key = 'PHigh'

    else:

        key = "Plow"
```

else:

if k < 0:

key = 'KHigh'

else:

key = "Klow"

response = Markup(str(fertilizer_dic[key]))

return render_template('fertilizer-result.html', recommendation=response,
title=title)

render disease prediction result page:

@app.route('/disease-predict', methods=['GET', 'POST'])

def disease_prediction():

title = 'Heck of Harvest - Disease Detection'

if request.method == 'POST':

if 'file' not in request.files:

return redirect(request.url)

file = request.files.get('file')

if not file:

return render_template('disease.html', title=title)

try:

img = file.read()

prediction = predict_image(img)

prediction = Markup(str(disease_dic[prediction]))

```
        return render_template('disease-result.html', prediction=prediction,
title=title)
```

```
    except:
```

```
        pass
```

```
    return render_template('disease.html', title=title)
```

```
if __name__ == '__main__':
```

```
    app.debug = True
```

```
    app.run()
```

```
a
```

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```
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)
```

8.TESTING

8.1 TEST CASES

Verify user is able to see the home page or not.

- Verify the UI elements in Home Page
- Verify user is able to redirect to predict page or not.
- Verify the UI elements in Predict Page
- Verify user is able to select the dropdown value or not.
- Verify user is able to upload the image or not.
- Verify whether the image is predicted correctly or not.

Test case ID	Feature Type	Component	Test Scenario	Pre-Requirement	Steps To Execute	Test Data	Expected Result	Actual Result	Status	Comments	TC Automation	ID
HomePage_TC_OO1	Functional	Home Page	Verify user is able to see the home page or not.		1. Enter URL and click go 2. verify whether the user is able to see the home page.	Enter URL and click go	User able to see the home page	Working as expected	Pass	Nil	N	-
HomePage_TC_OO2	UI	Home Page	Verify the UI elements in Home Page		1. Enter URL and click go 2. Verify the UI elements in Home Page.	Enter URL and click go	Application should show below UI elements: Home Tab & Predict Tab	Working as expected	pass	Nil	N	-
PredictPage_TC_OO3	Functional	Predict page	Verify user is able to redirect to predict page or not.		1. Enter URL and click go 2. Click on Predict button 3. Verify whether the user to redirect to predict page or not.	Click the predict button in home page	User should navigate to Predict page	Working as expected	pass	Nil	N	-
PredictPage_TC_OO4	UI	Predict page	Verify the UI elements in Predict Page		1. Enter URL and click go 2. Verify the UI elements in Predict Page.	Click the predict button and redirect to predict page	Application should show below UI elements: Dropdown List, Upload file Button, Predict button.	Working as expected	pass	Nil	N	-
PredictPage_TC_OO5	Functional	Predict page	Verify user is able to select the dropdown value or not.		1. Enter URL and click go 2. Click on Predict button 3. Verify whether the user to redirect to predict page or not. 4. Verify user is able to select the dropdown value or not.	Fruit or Vegetable	Application should shows user to choose fruit or vegetable option in dropdown list.	Working as expected	pass	Nil	N	-
PredictPage_TC_OO6	Functional	Predict page	Verify user is able to upload the image or not.		1. Enter URL and click go 2. Click on Predict button 3. Verify whether the user to redirect to predict page or not. 4. Verify user is able to select the dropdown value or not. 5. Verify user is able to upload the images or not	Images to be Uploaded	Application should shows the uploaded image.	Working as expected	pass	Nil	N	-
PredictPage_TC_OO7	Functional	Predict page	Verify whether the image is predicted correctly or not		1. Enter URL and click go 2. Click on Predict button 3. Verify whether the user to redirect to predict page or not. 4. Verify user is able to select the dropdown value or not. 5. Verify user is able to upload the images or not. 6. Verify whether the image is predicted correctly or not	Click the Predict Button	Application shows the predicted output	Working as expected	pass	Nil	N	-

8.2 USER ACCEPTANCE TESTING

Defect Analysis

This report shows the number of resolved or closed bugs at each severity level, and how they were resolved.

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
Leaf spots	14	4	2	3	19
Mosaic leaf pattern	13	6	3	6	28
Misshapen leaves	2	7	0	1	10
Yellow leaves	11	4	3	20	38
Fruit rots	3	2	1	0	6
Fruit spots	5	3	1	1	10
Blights	4	5	2	1	12
Totals	44	31	13	32	11

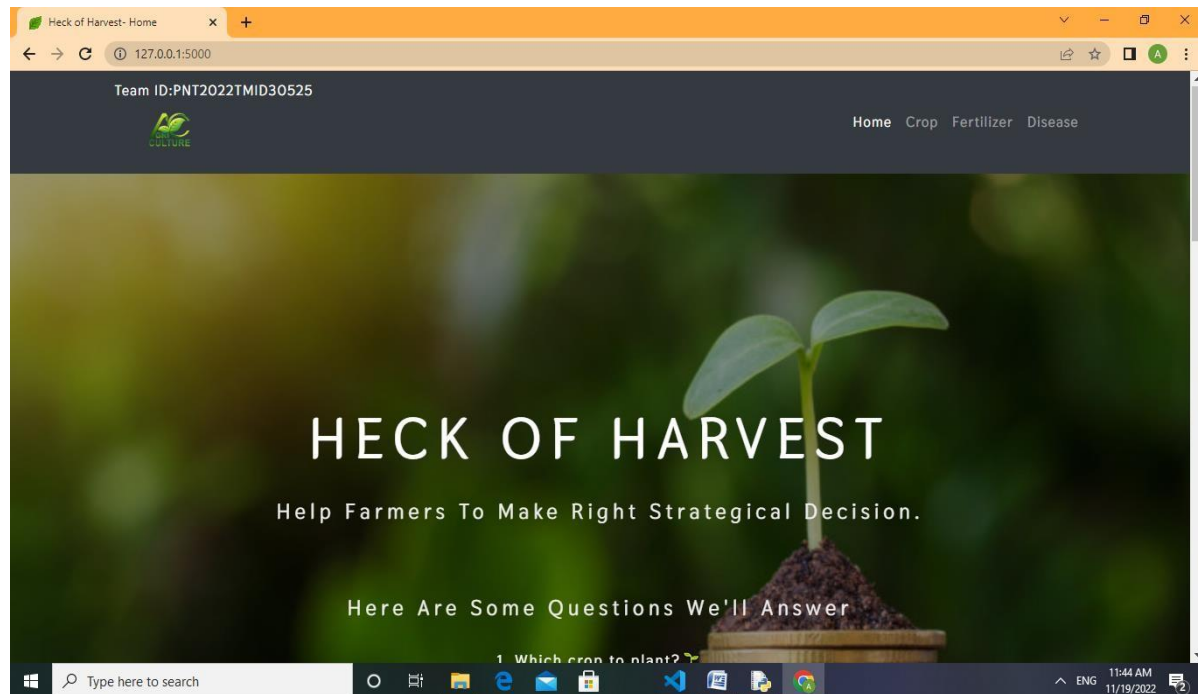
TEST CASE ANALYSIS

Test case analysis helps to understand the number of test case passed failed and untested.

section	Total Cases	Not Tested	Fail	Pass
Leaf spots	17	0	0	17
Mosaic leaf pattern	51	0	0	51
Misshapen leaves	20	0	0	20
Yellow leaves	7	0	0	7
Fruit rots	9	0	0	9
Fruit spots	4	0	0	4
Blights	2	0	0	2

9.RESULTS

HOME PAGE:



CROP RECOMMENDATION PAGE:

The screenshot shows the crop recommendation page of the 'Heck of Harvest' application. The browser address bar displays '127.0.0.1:5000/crop-recommend'. The page header includes 'Team ID:PNT2022TMID30525' and a navigation menu with 'Home', 'Crop', 'Fertilizer', and 'Disease'. The main content area has a light blue background and contains the text 'Find out the most suitable crop to grow in your farm'. Below this text are several input fields for agricultural data: 'Nitrogen' (with a placeholder 'Enter the value (example:50)'), 'Phosphorous' (with a placeholder 'Enter the value (example:50)'), 'Pottasium' (with a placeholder 'Enter the value (example:50)'), 'ph level' (with a placeholder 'Enter the value'), and 'Rainfall (in mm)' (with a placeholder 'Enter the value'). A yellow circular button with a right-pointing arrow is located at the bottom right of the input fields. The Windows taskbar at the bottom shows the search bar and various application icons.

FERTILIZER RECOMMENDATION PAGE:

The screenshot shows a web browser window with the URL `127.0.0.1:5000/fertilizer`. The page header includes the team ID `PNT2022TMID30525` and navigation links for `Home`, `Crop`, `Fertilizer`, and `Disease`. The main heading is "Provide suggestion for the crop growth based on the soil quality." Below this, there are four input fields: "Nitrogen" (with placeholder "Enter the value (example:50)"), "Phosphorous" (with placeholder "Enter the value (example:50)"), "Pottasium" (with placeholder "Enter the value (example:50)"), and "Crop you want to grow" (a dropdown menu with "Select crop" as the selected option). A blue "Predict" button is located at the bottom of the form. The browser's taskbar at the bottom shows the Windows search bar and various application icons, with the system clock indicating 11:51 AM on 11/19/2022.

DISEASE PREDICTION:

The screenshot shows a web browser window with the URL `127.0.0.1:5000/disease-predict`. The page header is identical to the first screenshot, showing the team ID `PNT2022TMID30525` and navigation links. The main heading is "Find out which disease has been caught by your plant". Below this, the text "Please Upload The Image" is displayed. There is a file upload interface with a "Choose File" button and the text "No file chosen". A blue "Predict" button is positioned below the upload area. The browser's taskbar at the bottom shows the Windows search bar and various application icons, with the system clock indicating 11:51 AM on 11/19/2022.

DISEASE PREDICTION RESULT:

The screenshot shows a web browser window with the URL `127.0.0.1:5000/disease-predict`. The page has a light blue background with a pink-to-blue gradient. The text on the page is as follows:

Crop: Pepper
Disease: Bacterial Spot

Cause of disease:

1. Bacterial spot is caused by several species of gram-negative bacteria in the genus *Xanthomonas*.
2. In culture, these bacteria produce yellow, mucoid colonies. A "mass" of bacteria can be observed oozing from a lesion by making a cross-sectional cut through a leaf lesion, placing the tissue in a droplet of water, placing a cover-slip over the sample, and examining it with a microscope (~200X)..

How to prevent/cure the disease

The browser's taskbar at the bottom shows the Windows search bar and various application icons. The system clock indicates 11:55 AM on 11/19/2022.

CROP RECOMMENDATION RESULT:

The screenshot shows a web browser window with the URL `127.0.0.1:5000/crop-predict`. The page has a dark blue header with the text "Team ID:PNT2022TMID30525" and a logo. The main content area is a light blue box with the following text:

It is recommended to grow *muskmelon* in your farm as it increase your yield and Productivity.

The browser's taskbar at the bottom shows the Windows search bar and various application icons. The system clock indicates 12:22 PM on 11/19/2022.

FERTILIZER RECOMMENDATION RESULT:

The screenshot shows a web browser window with the address bar displaying "127.0.0.1:5000/fertilizer-predict". The page content is displayed on a yellow background. At the top, it states "The K value of your soil is high." followed by "Please consider the following suggestions:". Below this, there is a list of six recommendations, each preceded by a number. The text is as follows:

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.

The browser window also shows a taskbar at the bottom with various application icons and a system clock indicating 11:56 AM on 11/19/2022.

9.1.PERFORMANCE METRICS

PERFORMANCE EVALUATION:

Performance testing is the practice of evaluating how a system performs in terms of responsiveness and stability under a particular workload. Performance tests are typically executed to examine speed, robustness, reliability, and application size.

The performance tests you run will help ensure your software meets the expected levels of service and provide a positive user experience. They will highlight improvements you should make to your applications relative to speed, stability, and scalability before they go into production. Applications released to the public in absence of testing might suffer from different types of problems that lead to a damaged brand reputation, in some cases, irrevocably.

The adoption, success, and productivity of applications depends directly on the proper implementation of performance testing.

While resolving production performance problems can be extremely expensive, the use of a continuous optimization performance testing strategy is key to the success of an effective overarching digital strategy.

Thus it helps in our project to detect the speed and accuracy of different data. In this we have deployed our project in IBM Watson cloud while the performance testing helps to evaluate the robustness and scalability.

S.N o.	Parameter	Values	Screenshot
1.	Model Summary	Total params:6,589,734 Trainable params: 6,589,734 Non-Trainable params:0	<pre> ===== Total params: 6,589,734 Trainable params: 6,589,734 Non-trainable params: 0 ----- Input size (MB): 0.75 Forward/backward pass size (MB): 343.95 Params size (MB): 25.14 Estimated Total Size (MB): 369.83 ----- None </pre>
2.	Accuracy	Training Accuracy – 83% Validation Accuracy -99%	<pre> Epoch [0], last_lr: 0.00012, train_loss: 0.7451, val_loss: 0.5340, val_acc: 0.8330 Epoch [1], last_lr: 0.00000, train_loss: 0.1219, val_loss: 0.0256, val_acc: 0.9922 CPU times: user 15min 13s, sys: 15min 22s, total: 30min 36s Wall time: 32min 10s </pre>

ACCURACY:

```

Epoch [0], last_lr: 0.00012, train_loss: 0.7451, val_loss: 0.5340, val_acc: 0.8330
Epoch [1], last_lr: 0.00000, train_loss: 0.1219, val_loss: 0.0256, val_acc: 0.9922
CPU times: user 15min 13s, sys: 15min 22s, total: 30min 36s
Wall time: 32min 10s

```

MODEL SUMMARY:

```
print(summary(model.cuda(), (INPUT_SHAPE)))
```

Layer (type)	Output Shape	Param #
Conv2d-1	[-1, 64, 256, 256]	1,792
BatchNorm2d-2	[-1, 64, 256, 256]	128
ReLU-3	[-1, 64, 256, 256]	0
Conv2d-4	[-1, 128, 256, 256]	73,856
BatchNorm2d-5	[-1, 128, 256, 256]	256
ReLU-6	[-1, 128, 256, 256]	0
MaxPool2d-7	[-1, 128, 64, 64]	0
Conv2d-8	[-1, 128, 64, 64]	147,584
BatchNorm2d-9	[-1, 128, 64, 64]	256
ReLU-10	[-1, 128, 64, 64]	0
Conv2d-11	[-1, 128, 64, 64]	147,584
BatchNorm2d-12	[-1, 128, 64, 64]	256
ReLU-13	[-1, 128, 64, 64]	0
Conv2d-14	[-1, 256, 64, 64]	295,168
BatchNorm2d-15	[-1, 256, 64, 64]	512
ReLU-16	[-1, 256, 64, 64]	0
MaxPool2d-17	[-1, 256, 16, 16]	0
Conv2d-18	[-1, 512, 16, 16]	1,180,160
BatchNorm2d-19	[-1, 512, 16, 16]	1,024
ReLU-20	[-1, 512, 16, 16]	0
MaxPool2d-21	[-1, 512, 4, 4]	0
Conv2d-22	[-1, 512, 4, 4]	2,359,808
BatchNorm2d-23	[-1, 512, 4, 4]	1,024
ReLU-24	[-1, 512, 4, 4]	0
Conv2d-25	[-1, 512, 4, 4]	2,359,808
BatchNorm2d-26	[-1, 512, 4, 4]	1,024
ReLU-27	[-1, 512, 4, 4]	0
MaxPool2d-28	[-1, 512, 1, 1]	0
Flatten-29	[-1, 512]	0
Linear-30	[-1, 38]	19,494

ADVANTAGES:

The system comes with a model to be precise and accurate in predicting crop yield and deliver the end user with proper recommendations about required fertilizer ratio based on atmospheric and soil parameters of the land which enhance to increase the crop yield and increase farmer revenue. The prediction of crop yield based on location and proper implementation of algorithms have proved that the higher crop yield can be achieved. From above work I conclude that for soil classification Random Forest is good with accuracy 86.35% compare to Support Vector Machine. For crop yield prediction Support Vector Machine is good with accuracy 99.47% compare to Random Forest algorithm. The work can be extended further to add following functionality. Mobile application can be build to help farmers by uploading image of farms. Crop diseases detection using image processing in which user get pesticides based on disease images. Implement Smart Irrigation System for farms to get higher yield.

- Fertilizers have all nutrients required for plants growth.
- It is soluble and easily absorbed by plants.
- It enhances the metabolism of plants.
- It is easily available in the market.
- Highly needed for large production.

DISADVANTAGES:

- Fertilizers are more expensive than manure.
- Over fertilization can damage the plants.
- It is toxic and can harm humans.
- It affected the environment and echo system.
- Long term use reduce soil quality.

CONCLUSIONS:

The core strategy of this project is to predict the crop based on the soil nutrient content and the location where the crop is growing. This system will help the farmers to choose the right crop for their land and to give the suitable amount of fertilizer to produce the maximum yield. The Support Vector Machine algorithm helps to predict the crop precisely based on the pre-processed crop data. This system will also help the new comers to choose the crop which will grow in their area and produce them a good profit.

A decent amount of profit will attract more people towards the agriculture. Also, the crop growth is based on the climate conditions in the particular area and the seasonal monsoons happens now are unpredictable, hence it is easy for the farmers when the prediction result is also based on the climatic conditions. Live weather prediction will also help the users to predict the crop water needs and also it will help the farmers to decrease the crop damage due to the rain or drought.

The prediction of crop yield based on soil data and proper implementation of algorithms have proved that a higher crop yield can be achieved. From the above work, we conclude that for soil classification Random Forest is a suitable algorithm with an accuracy of 99.09% compare to Gaussian Naive Bayes. The work can be extended further to add the following functionality. Building a Website can be built to help farmers by uploading an image of farms. Crop diseases detection uses image processing in which users get pesticides based on disease images and Fertilizer prediction based on soil condition.

By categorizing the soil samples according to the soil type, land type and macro nutrients Nitrogen (N), Phosphorus (P) and Potassium (K) present in the soil the suitable crop along with its appropriate fertilizer is suggested to the agricultural stakeholder. The month in which the yield will be high is also suggested to the user. The yield calculation is also provided for the crop selected by the farmer. The proposed crop recommendation system provides 82% of accuracy.

FUTURE SCOPE:

The future work is to implement Machine Learning Algorithms like Ensemble Classifiers to predict the crop yield and recommend the crop with appropriate fertilizer. In the existing system only soil characteristics were considered to provide crop recommendations. In the future work the climatic parameters will also be taken into account to provide crop recommendations. Also the method can be extended to include diverse varieties of crop to be cultivated and to analyze its performance.

This further research is implementing the proposed algorithm with the existing public datasets. Also, various segmentation algorithms can be implemented to improve accuracy. The proposed algorithm can be modified further to identify the disease that affects the various plant organs such as stems and fruits.

DEMO LINK:

https://drive.google.com/file/d/1fbkuCAV9rl2a4oknHqLrLvD3O5qsXd3Z/view?usp=share_link