FERTILIZERS RECOMMENDATION SYSTEM FOR DISEASE PREDICTION

TEAM ID: PNT2022TMID08559

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1. INTRODUCTION

1.1 PROJECT OVERVIEW

Agriculture serves as a means of supplying food to a population that is always expanding, as well as a significant source of energy and a means of combating global warming. Plant diseases are very important because they can have a negative impact on the quality and quantity of crops produced in agriculture. Early detection of plant diseases is crucial for their treatment and management. Typically, illnesses are identified using the naked eye technique. Experts who can recognise variations in leaf colour are involved in this process. This method requires a lot of work, takes a while, and is not appropriate for fields with a lot of space. The same ailment is frequently classified differently by various experts. Costly specialist monitoring is required for this procedure, which makes it pricey. Plant diseases can drive up the cost of agricultural production and, if left untreated at an early stage, could spell complete financial ruin for a producer. In order to stop the spread of a plant disease at a low cost and save the majority of the production, farmers must keep an eye on their crops and recognise the first symptoms. It may be expensive to hire experienced agriculturists, particularly in remote, isolated geographic areas .Various experts regularly assign multiple classifications to the same illness. This operation is costly because it calls for pricy professional supervision. Plant diseases can increase the cost of agricultural production and, if not promptly treated, could result in a producer's total financial ruin. Farmers must keep an eye on their crops and be able to spot the first symptoms in order to stop the spread of a plant disease at a low cost and save the majority of the production. Agriculturists with experience may be expensive to hire, especially in isolated, distant locations.

1.2 PURPOSE

They forecast plant disease and recommend fertiliser for the damaged plants. This frequently involves a range of methods for assessing the qualities of the herbs that largely influence the plants. These complex systems that contain a large amount of datasets are forecasted using the neural network. By using artificial intelligence, complex manual systems' working models can be made simpler and more precise.

2. LITERATURE SURVEY

2.1 EXISTING PROBLEM

Leaves are the most obvious and widespread choice for tree species recognition, even though the botanical classification was not built upon their properties. They can be found almost all year long, are easy to photograph, and their shapes present well studied specificities that make the identification, if not trivial, possible. Our goal with the Folia application is then to build a system for leaf shape analysis that processes, unlike what has been done to date, pictures in a natural environment. With the aim of being an educational tool, it relies on highlevel geometric criteria inspired by those used by botanists, that make a semantic interpretation possible, to classify a leaf into a list of species. Digital image processing will improve the quality of the image by removing noise & other unwanted pixels and obtain more information from image. Image segmentation is a mid-level processing technique used to analyze the image and can be used to classify or cluster an image into several disjoint parts by grouping the pixels to form a region of homogeneity based on the pixel characteristics like gray level, color, texture, intensity and other features. The main purpose of the segmentation process is to get more information about the image, the region we are interested in and to clearly differentiate the object and the background in an image. The criteria for segmenting the image is very hard to decide as it varies from image to image and also varies significantly on the modal quality of image. In some cases interactive methods can be laborious and time

consuming and in some cases manual interaction to segment the image may be error-prone while the fully automated approach can give error output.

2.1.1LIMITATIONS OF EXISTING SYSTEM:

- Suffer in the local minima problems.
- Dimensionality is high to produce large number of irrelevant features.
- User defined segmentation can be done.
- Does not recommend the fertilizers to leaves diseases.

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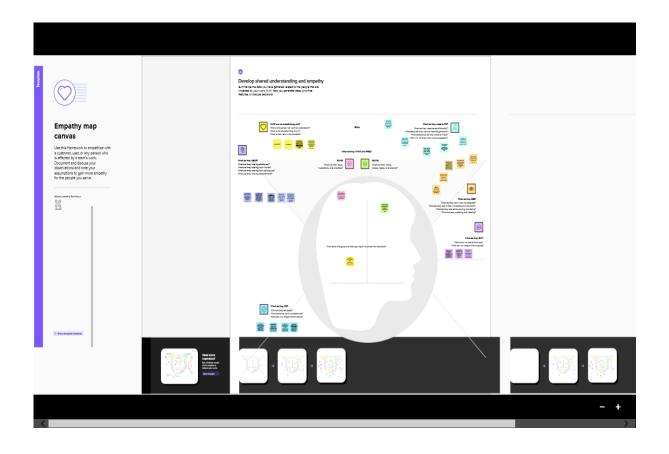
2.3 PROBLEM STATEMENT DEFINITION

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.

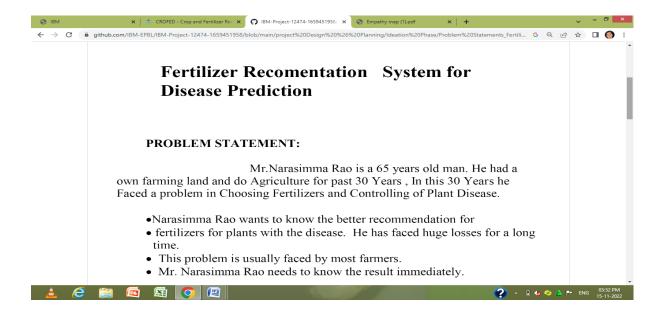
3. IDEATION & PROPOSED SOLUTION

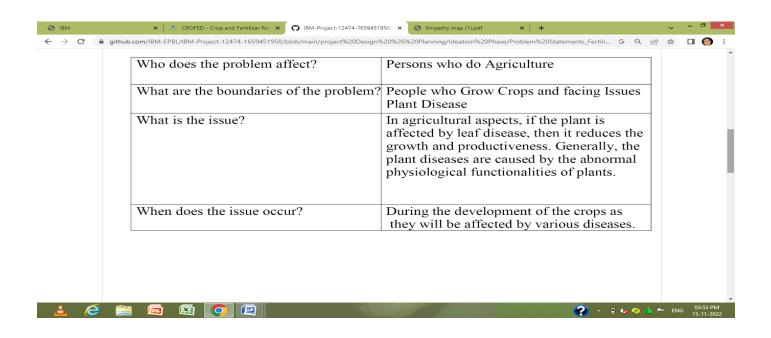
3.1 EMPATHY MAP CANVAS

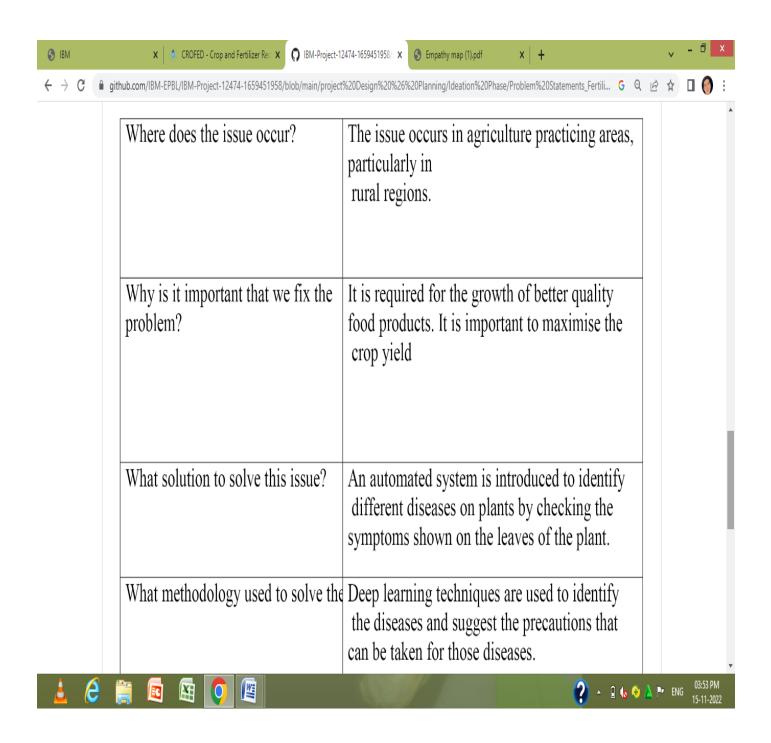


3.2 IDEATION & BRAINSTORMING

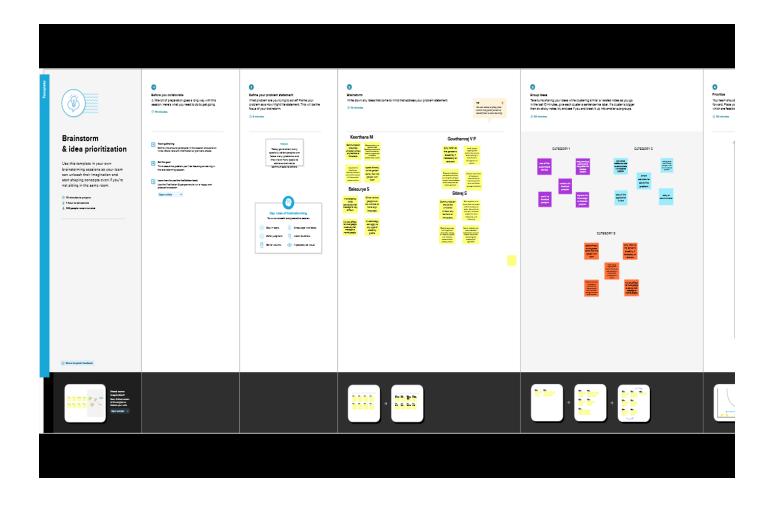
3.2.1 IDEATION:

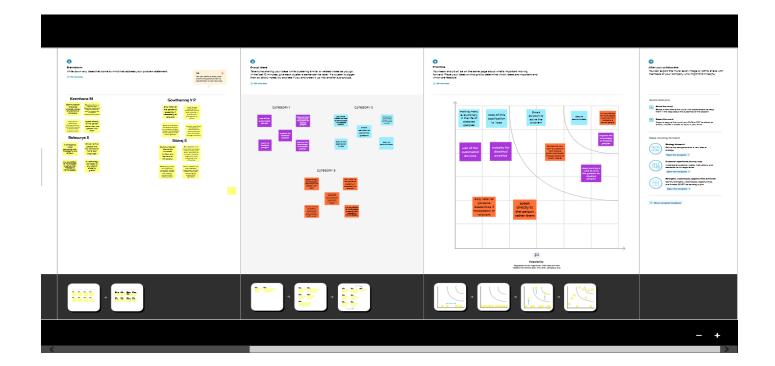






3.2.2 BRAINSTORMING:





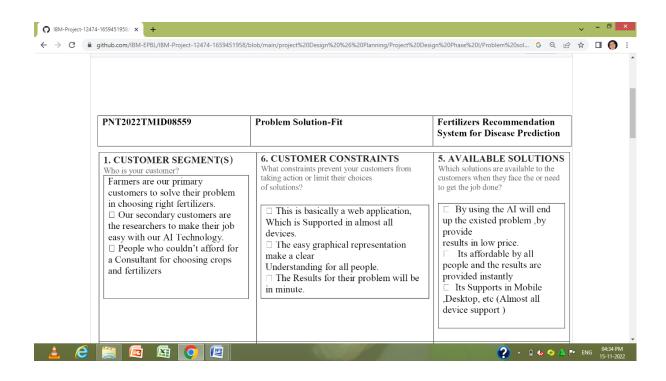
3.3 PROPOSED SOLUTION:

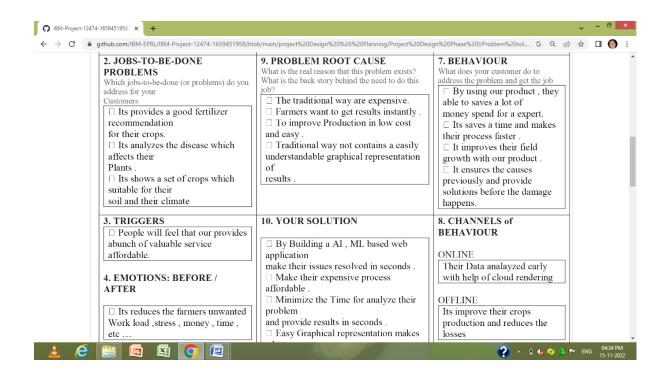
Even when considering trees only, leaves show an impressively wide variety in shapes. It is however necessary to come up with a representation of what a leaf is, that is accurate enough to be fitted to basically any kind of leaf. The general shape of a leaf is a key component of the process of identifying a leaf. Botanists have a whole set of terms describing either the shape of a simple leaf, of the lobes of a palmate leaf, or of the leaflets of a compound leaf. Here present a study on segmentation of leaf images restricted to semi-controlled conditions, in which leaves are photographed against a solid light-colored background. Such images can be used in practice for plant species identification, by analyzing the distinctive shapes of the leaves. The most important of these are: the variety of leaf shapes, inevitable presence of shadows and specularities, and the time constraints required by interactive species identification applications. The identification of species is the first and essential key to understand the plant environment. In this project introduce a method designed to deal with the obstacles raised by such complex images, for simple and lobed tree leaves. A first segmentation step based on a light polygonal leaf model is first performed, and later used to guide the evolution of an active contour. Combining global shape descriptors given by the polygonal model with local curvature-based features, the leaves are then classified over leaf datasets. In this project we introduce a method designed to deal with the obstacles raised by such complex images, for simple and lobed tree leaves. A first segmentation step based on graph cut approach is first performed, and later used to guide the evolution of leaf boundaries. And implement classification algorithm to classify the diseases and recommend the fertilizers to affected leaves.

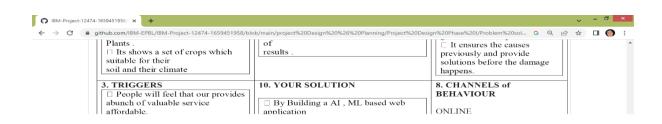
3.3.1ADVANTAGE OF PURPOSED SYSTEM:

- Segmentation can be done easily to spilt the tree parts.
- Classify the affected parts in leaves.
- Eliminate redundant features of images.
- Provide improved accuracy rate.

3.4 PROBLEM SOLUTION FIT:







4. REQUIREMENT ANALYSIS

REQUIREMENT SPECIFICATION

The technical specification requirement for the software products is the requirement specification. It lists the functional, performance, and security requirements for a specific software system. Additionally, usage scenarios from a user, operational, and administrative standpoint are provided in the requirements. A thorough overview of the software project is what the software requirements specification is meant to do. The target audience is given a description of the project's parameters, goals, user interface, hardware, and software needs.

4.1 FUNCTIONAL REQUIREMENT:

Operating system : Windows OS

• Front End : C#.NET

Back End : SQL SERVER

• Application : Windows Application

• Tool : Visual Studio 2010

4.2 NON-FUNCTIONAL REQUIREMENT:

• Processor : Dual core processor 2.6.0 GHZ

• RAM : 2GB

Hard disk : 160 GBCompact Disk : 650 Mb

Keyboard : Standard keyboard

• Monitor : 15 inch color monitor

5. PROJECT DESIGN

5.1 DATA FLOW DIAGRAM:

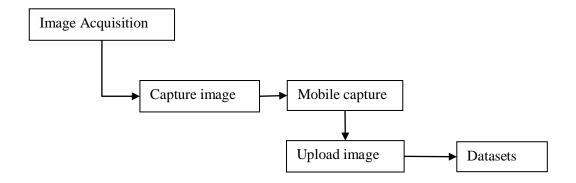
FLOW DIAGRAM

A two-dimensional diagram explains how data is processed and transferred in a system. The graphical depiction identifies each source of data and how it interacts with other data sources to reach a common output. Individuals seeking to draft a data flow diagram must identify external inputs and outputs, determine how the inputs and outputs relate to each other, and explain with graphics how these connections relate and what they result in. This type of diagram helps business development and design teams visualize how data is processed and identify or improve certain aspects.

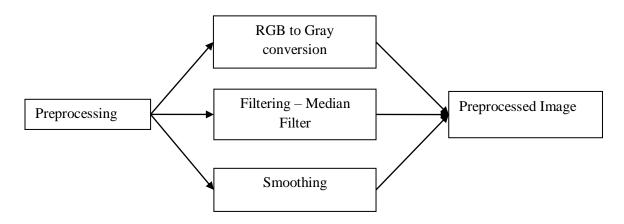
Data flow Symbols:

Symbol	Description
	An entity . A source of data or a destination for data.
	A process or task that is performed by the system.
	A data store , a place where data is held between processes.
	A data flow.

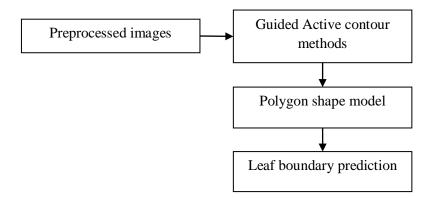
Level 0



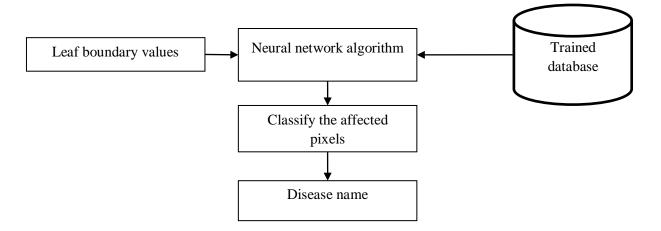
Level 1



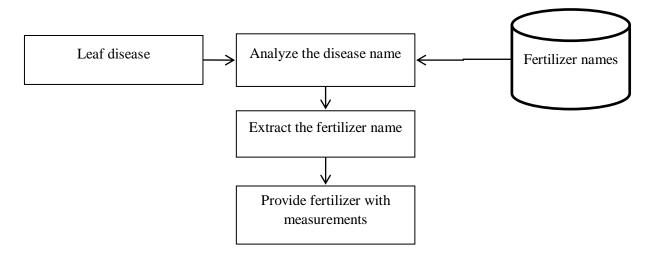
Level 2



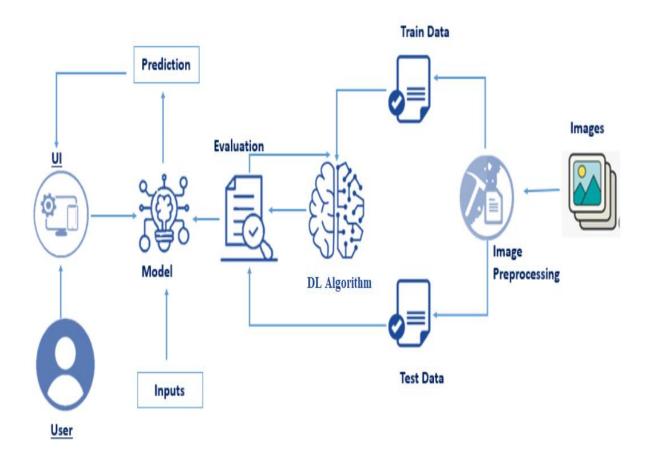
Level 3



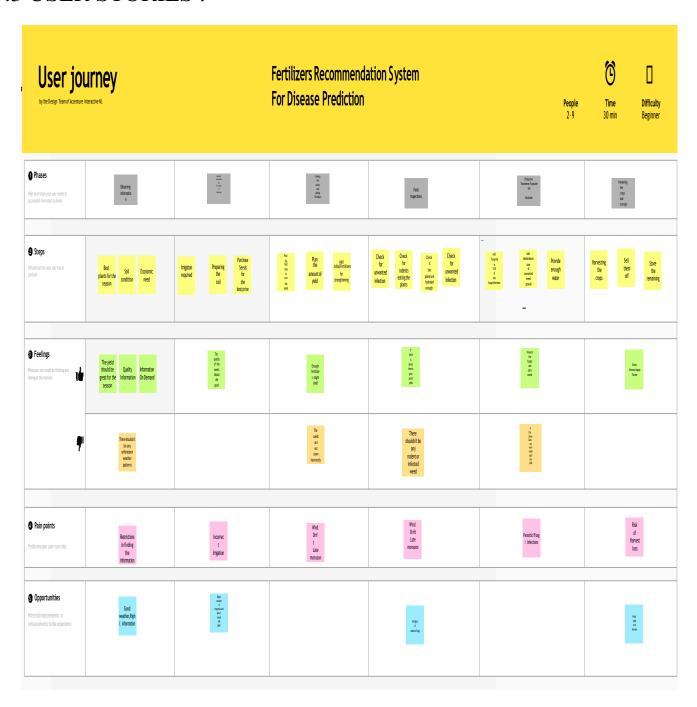
Level 4



5.2 SOLUTION & TECHNICAL ARCHITECTURE :



5.3 USER STORIES:



6. PROJECT PLANNING & SCHEDULING

6.1 SPRINT PLANNING & ESTIMATION:

PROJECT PLANNING PHASE MILESTONE & ACTIVITY LIST

DATE	9 NOVEMBER 2022
TEAM ID	PNT2022TMID08559
PROJECT NAME	FERTILIZER RECOMMENDATION SYSTEM FOR PLANT DISEASEPREDICTION

Milestone:

Modern Technology are increasing and optimizing the Performance of the Artificial Intelligences (AI) Model. Based Crop Yield Disease Prediction System, is helpful for farmersto prevent the crop from the various Disease which can identify the Disease with in a processof capturing the Image at the plant and Machine Learning Algorithm will give affected Disease Name. In this Project Milestone will be given the Best Solution for the farmer using the complete friendly and simple user interface web application to fetching the solution by own. In addition, process we are planned to add a valid Module that is Fertilizer recommendation for the Specific Disease. It can give both artificial fertilizer and Natural Fertilizer in suggestion manner.

Activity List:

In Project Management Planning is an important task to scheduling the phases of the project to the Team Member. In this Activity can shows the various activity are allocated and doneby the Team Members. The phases are

Phase 1: Literature survey and information

gathering

Phase 2: Prepare empathy map

Phase 3: Ideation

Phase 4: Proposed solution

Phase 5: Proposed solution fit

Phase 6: Solution architecture

Phase 7: Customer journey

Phase 8: Functional requirement

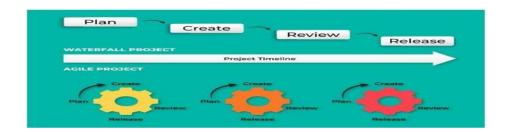
Phase 9: Data flow diagrams

Phase 10: Technology architecture

Phase 11: prepare milestone activity and list

Phase 12: Spirit delivery plan

Phase 13: Project development-Delivery



6.2 SPRINT DELIVERY SCHEDULE:

Project Planning Phase Project Planning Template (Product Backlog, Sprint Planning, Stories, Story points)

DATE	10 NOVEMBER
TEAM ID	PNT2022TMIDO8559
PROJECT TITLE	Fertilizers Recommendation System For Disease Prediction

Product Backlog, Sprint Schedule, and Estimation:

Sprin t	Function Requirement (Epic)	User Story · Numbe r	User Story / Task	Story Points (Total)	Priorit y	Team Members
Sprint-1	Model Creation and Training (Fruits)		Create a model which can classify diseased fruit plants from given images. I also need to test the model and deploy it on IBM Cloud	8	High	Vishwanantha, sathishkumar, deva, jaya raj
	Model Creation and Training (Vegetables)		Create a model which can classify diseased vegetable plants from given images	2	High	Vishwanantha, sathishkumar, deva, jaya raj

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points (Total)	Priorit y	Team Members
Sprint-2	Model Creation and Training (Vegetables)		Create a model which can classify diseased vegetable plants from given images and train on IBM Cloud	6	high	Vishwanantha, sathishkumar, deva, jaya raj
	Registration	USN-1	As a user, I can register by entering my email, password, and confirming my password or via O Auth API	3	high	Vishwanantha, sathishkumar, deva, jaya raj
	Upload page	USN-2	As a user, I will be redirected to a page where I can upload my pictures of crops	4	high	Vishwanantha, sathishkumar, deva, jaya raj
	Suggestion results	USN-3	As a user, I can view the results and then obtain the suggestions provided by the ML mode	4	high	Vishwanantha, sathishkumar, deva, jaya raj
	Base Flask App		A base Flask web app must be created as an interface for the ML model	2	high	Vishwanantha, sathishkumar, deva, jaya raj

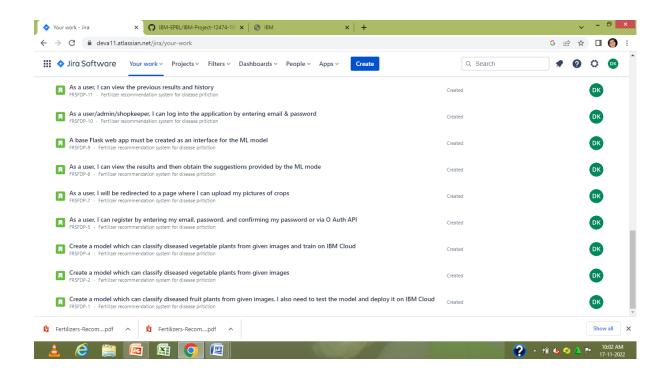
Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points (Total)	Priority	Team Members
Sprint-3	Login	USN-4	As a user/admin/shopkeeper, I can log into the application by entering email & password	2	high	Vishwanantha, sathishkumar, deva, jaya raj
	User Dashboard	USN-5	As a user, I can view the previous results and history	3	Medium	Vishwanantha, sathishkumar, deva, jaya raj
	Integration		Integrate Flask, CNN model with Cloud ant DB	5	Medium	Vishwanantha, sathishkumar, deva, jaya raj
	Containerization		Containerize Flask app using Docker	2	low	Vishwanantha, sathishkumar, deva, jaya raj

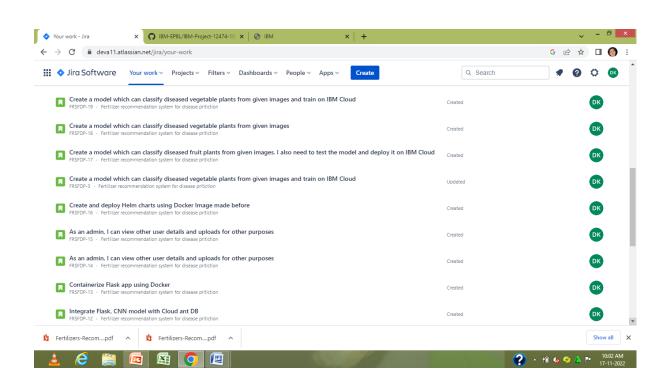
Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points (Total)	Priority	Team Members
Sprint-4	Dashboard (Admin)	USN-6	As an admin, I can view other user details and uploads for other purposes	2	Medium	Vishwanantha, sathishkumar, deva, jaya raj
	Dashboard (Shopkeeper)	USN-7	As a shopkeeper, I can enter fertilizer products and then update the details if any	2	low	Vishwanantha, sathishkumar, deva, jaya raj
	Containerization		Create and deploy Helm charts using Docker Image made before	2	low	Vishwanantha, sathishkumar, deva, jaya raj

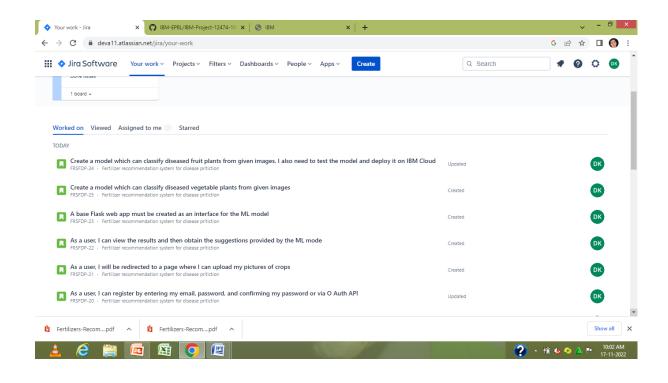
Project Tracker, Velocity & Burn down Chart:

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	06 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

6.3 REPORTS FROM JIRA:







7. CODING & SOLUTIONING

7.1 FEATURE 1: CODING SAMPLE

```
mport tensorflow as tf
import numpy as np

from tkinter import *
import os
from tkinter import filedialog
import cv2
import time
from matplotlib import pyplot as plt
from tkinter import messagebox

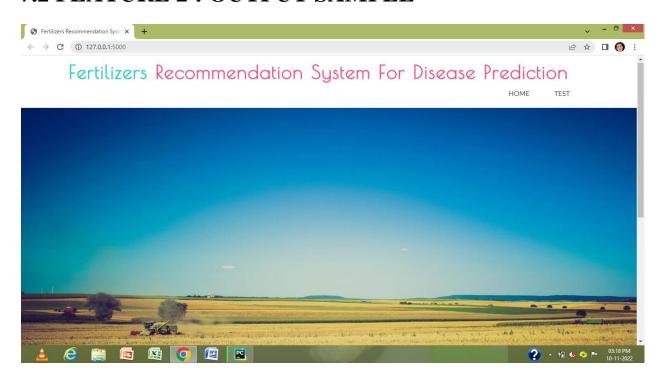
def endprogram():
    print ("\nProgram terminated!")
    sys.exit()
```

```
def fulltraining():
    import model as mm
def testing():
    global testing_screen
    testing_screen = Toplevel(main_screen)
    testing screen.title("Testing")
    # login screen.geometry("400x300")
    testing_screen.geometry("600x450+650+150")
    testing_screen.minsize(120, 1)
    testing_screen.maxsize(1604, 881)
    testing screen.resizable(1, 1)
    testing_screen.configure(bg='green')
    Label(testing_screen, text='''Upload Image''', disabledforeground="#a3a3a3",
          foreground="#000000", width="300", height="2", font=("Calibri",
16)).pack()
    Label(testing_screen, text="").pack()
    Label(testing screen, text="").pack()
    Label(testing_screen, text="").pack()
    Button(testing_screen, text='''Upload Image''', font=(
        'Verdana', 15), height="2", width="30", command=imgtest).pack()
global affect
def imgtest():
    import_file_path = filedialog.askopenfilename()
    image = cv2.imread(import file path)
    print(import_file_path)
    filename = 'Output/Out/Test.jpg'
    cv2.imwrite(filename, image)
    print("After saving image:
```

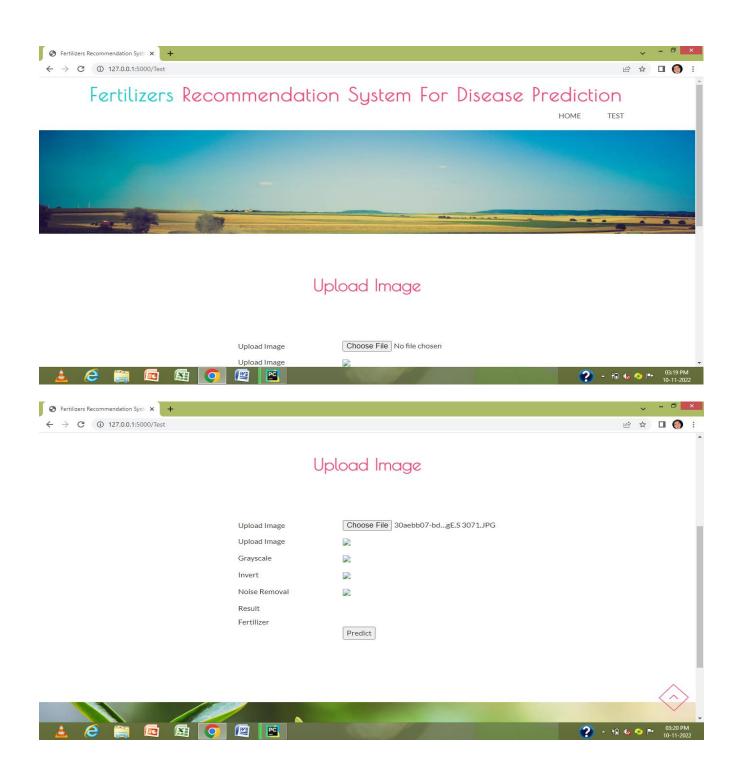
```
#import_file_path = filedialog.askopenfilename()
   print(import_file_path)
   fnm = os.path.basename(import_file_path)
   print(os.path.basename(import_file_path))
  # file sucess()
   img = cv2.imread(import_file_path)
   if img is None:
       print('no data')
   img1 = cv2.imread(import_file_path)
   print(img.shape)
   img = cv2.resize(img, ((int)(img.shape[1] / 5), (int)(img.shape[0] / 5)))
   original = img.copy()
   neworiginal = img.copy()
   cv2.imshow('original', img1)
   gray = cv2.cvtColor(img1, cv2.COLOR_BGR2GRAY)
   img1S = cv2.resize(img1, (960, 540))
   cv2.imshow('Original image', img1S)
   grayS = cv2.resize(gray, (960, 540))
   cv2.imshow('Gray image', grayS)
   dst = cv2.fastNlMeansDenoisingColored(img1, None, 10, 10, 7, 21)
   cv2.imshow("Nosie Removal", dst)
   thresh = 127
   im_bw = cv2.threshold(grayS, thresh, 255, cv2.THRESH_BINARY)[1]
   #cv2.imshow("affect Removal", im_bw)
number_of_black_pix = np.sum(im_bw == 0)
   #print(number_of_black_pix)
       #affect =
   result()
def result():
   import warnings
   warnings.filterwarnings('ignore')
   import tensorflow as tf
   classifierLoad = tf.keras.models.load model('firemodel.h5')
   import numpy as np
   from keras.preprocessing import image
   test_image = image.load_img('Output/Out/Test.jpg', target_size=(200, 200))
```

```
img1 = cv2.imread('Output/Out/Test.jpg')
    # test_image = image.img_to_array(test_image)
    test_image = np.expand_dims(test_image, axis=0)
    result = classifierLoad.predict(test_image)
    pre=''
    if result[0][0] == 1:
        out="Fire"
    elif result[0][1] == 1:
        out="Nofire"
    messagebox.showinfo("Result", "Classfication Rssult : "+str(out))
def main_account_screen():
    global main_screen
    main_screen = Tk()
    width = 600
    height = 600
    screen_width = main_screen.winfo_screenwidth()
    screen_height = main_screen.winfo_screenheight()
    x = (screen_width / 2) - (width / 2)
y = (screen_height / 2) - (height / 2)
main_screen.geometry("%dx%d+%d+%d" % (width, height, x, y))
    main_screen.resizable(0, 0)
# main_screen.geometry("300x250")
    main_screen.configure(bg='green')
    main screen.title("Forest Fire Detection ")
    Label(text="Forest Fire Detection", width="300", height="5", bg='green',
font=("Calibri", 16)).pack()
    Button(text="Training", font=(
         'Verdana', 15), height="2", width="30", bg='green',command=fulltraining,
highlightcolor="black").pack(side=TOP)
    Label(text="").pack()
    Button(text="Testing", font=(
        'Verdana', 15), height="2", width="30",bg='green',
command=testing).pack(side=TOP)
    Label(text="").pack()
    main_screen.mainloop()
```

7.2 FEATURE 2: OUTPUT SAMPLE

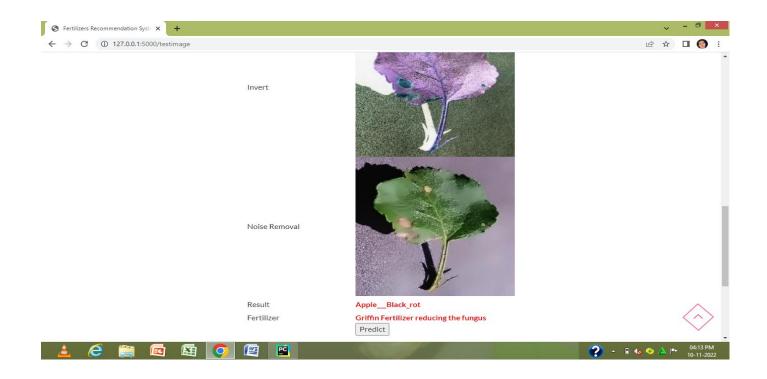












8. TESTING

8.1 TEST CASE:

8.1 TEST CASES

A test case has components that describe input, action and an expected response, in order to determine if a feature of an application is working correctly. A test case is a set of instructions on "HOW" to validate a particular test objective/target, which when followed will tell us if the expected behavior of the system is satisfied or not.

Characteristics of a good test case:

Accurate: Exacts the purpose.

Economical: No unnecessary steps or words.

• Traceable: Capable of being traced to requirements.

• Repeatable: Can be used to perform the test over and over.

S.NO	FUNCTION	DESCRIPTION	EXPECTED	ACTUAL	STATUS
			OUTPUT	OUTPUT	
1	Framework	Generate the	Individual	Individual	Success
	construction	GUI for admin	page for	page for	
		and user	admin and	admin and	
			user	user	
2	Read the	Comments	Comments in	Comments	Success
	comments	analysis	text format	in text	
				format	
3	Classification	Classify the	Negative	Negative	Success
		datasets	comments	comments	
4	Rules	Block the	Block the	Block the	Success
	implementation	comments and	users	users	
		friends			

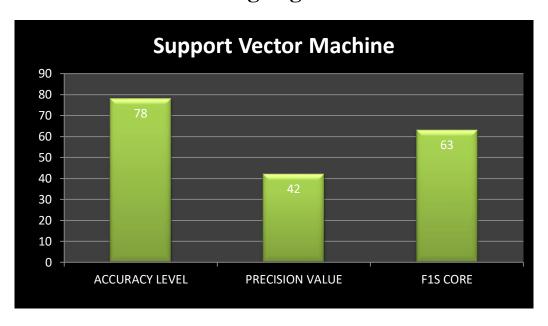
8.2 USER ACCEPTANCE TESTING:

Acceptance testing can be defined in many ways, but a simple definition is the succeeds when the software functions in a manner that can be reasonable expected by the customer. After the acceptance test has been conducted, one of the two possible conditions exists. This is to fine whether the inputs are accepted by the database or other validations. For example accept only numbers in the numeric field, date format data in the date field. Also the null check for the not null fields. If any error occurs then show the error messages. The function of performance characteristics to specification and is accepted. A deviation from specification is uncovered and a deficiency list is created. User Acceptance Testing is a critical phase of any project and requires significant participation by the end user. It also ensures that the system meets the functional requirements.

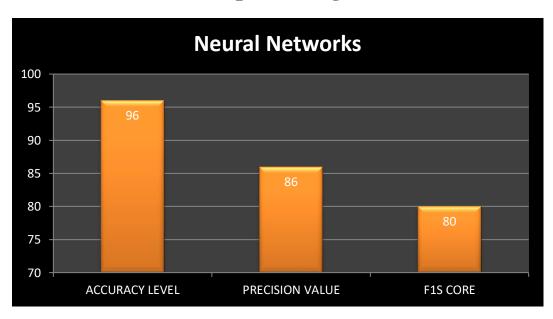
9. RESULT

9.1 PERFORMANCE METRICS

Existing Algorithm



Proposed Algorithm



10. ADVANTAGES AND DISADVANTAGES

DISADVANTAGES

- Suffer in the local minima problems.
- Dimensionality is high to produce large number of irrelevant features.
- User defined segmentation can be done.
- Does not recommend the fertilizers to leaves diseases.
- Manual approach is used

ADVANTAGES

- Segmentation can be done easily to spilt the tree parts.
- Classify the affected parts in leaves.
- Eliminate redundant features of images.
- Provide improved accuracy rate

11. CONCLUSION

CONCLUSION

We presented a machine learning approach for crop yield prediction, which demonstrated superior performance in Crop Challenge using large datasets of products. The approach used deep neural networks to make yield predictions (including yield, check yield, and yield difference) based on genotype and environment data. The carefully designed deep neural networks were able to learn nonlinear and complex relationships between genes, environmental conditions, as well as their interactions from historical data and make reasonably accurate predictions of yields for new hybrids planted in new locations with known weather conditions. Performance of the model was found to be relatively sensitive to the quality of weather prediction, which suggested the importance of weather prediction techniques. We trained two deep neural networks, one for yield and the other for check

yield, and then used the difference of their outputs as the prediction for yield difference. This model structure was found to be more effective than using one single neural network for yield difference, because the genotype and environment effects are more directly related to the yield and check yield than their difference. In modern era, the deep neural network is the prominent tool in agricultural industry for providing support to farmers in monitoring crop yield based on multiple parameters. Thus, the machine learning model provides high accuracy in detecting the suitable crop identification compared to other methodologies.

12. FUTURE SCOPE

FUTURE WORK

This project describes crop yield prediction ability of the algorithm. In future we can determine the efficient algorithm based on their accuracy metrics that will helps to choose an efficient algorithm for crop yield prediction

13. APPENDIX

SOURCE CODE:

```
import tensorflow as tf
import time
import numpy as np
import os

start = time.time()
#try:
# Total iterations
final_iter = 1000

# Assign the batch value
batch_size = 20

# 20% of the data will automatically be used for validation
validation_size = 0.2
img_size = 128
num_channels = 3
```

```
train_path = r'data\Train'
# Prepare input data
if not os.path.exists(train_path):
print("No such directory")
raise Exception
classes = os.listdir(train_path)
num classes = len(classes)
# We shall load all the training and validation images and labels into memory
using openCV and use that during training
data = dataset.read_train_sets(train_path, img_size, classes,
validation size=validation size)
# Display the stats
print("Complete reading input data. Will Now print a snippet of it")
print("Number of files in Training-set:\t\t{}".format(len(data.train.labels)))
print("Number of files in Validation-set:\t{}".format(len(data.valid.labels)))
session = tf.compat.v1.Session()
x = tf.compat.v1.placeholder(tf.float32, shape=[None, img_size, img_size,
num channels], name='x')
## LabeLs
y true = tf.compat.v1.placeholder(tf.float32, shape=[None, num classes],
name='y_true')
y true cls = tf.argmax(y true, dimension=1)
##Network graph params
filter size conv1 = 3
num filters conv1 = 32
filter size conv2 = 3
num_filters_conv2 = 32
filter size conv3 = 3
num filters conv3 = 64
fc layer size = 128
def create weights(shape):
return tf.Variable(tf.random.truncated normal(shape, stddev=0.05))
def create biases(size):
return tf.Variable(tf.constant(0.05, shape=[size]))
def make generator model(input,
                               num_input_channels,
                               conv filter size,
                               num filters):
## We shall define the weights that will be trained using create weights function.
weights = create weights(shape=[conv filter size, conv filter size,
num input channels, num filters])
## We create biases using the create_biases function. These are also trained.
biases = create_biases(num_filters)
```

```
## Creating the convolutional layer
layer = tf.nn.conv2d(input=input,
filter=weights,
strides=[1, 1, 1, 1],
padding='SAME')
    layer += biases
## We shall be using max-pooling.
layer = tf.nn.max pool(value=layer,
ksize=[1, 2, 2, 1],
strides=[1, 2, 2, 1],
padding='SAME')
## Output of pooling is fed to Relu which is the activation function for us.
layer = tf.nn.relu(layer)
return layer
# Function to create a Flatten Layer
def create flatten layer(layer):
# We know that the shape of the layer will be [batch_size img_size img size
num channels]
    # But let's get it from the previous layer.
layer_shape = layer.get_shape()
## Number of features will be img_height * img_width* num_channels. But we shall
calculate it in place of hard-coding it.
num_features = layer_shape[1:4].num_elements()
## Now, we Flatten the layer so we shall have to reshape to num features
layer = tf.reshape(layer, [-1, num features])
return layer
# Function to create a Fully - Connected Layer
def create_fc_layer(input,
                    num_inputs,
                    num outputs,
                    use_relu=True):
# Let's define trainable weights and biases.
weights = create weights(shape=[num inputs, num outputs])
    biases = create_biases(num_outputs)
# Fully connected layer takes input x and produces wx+b.Since, these are matrices,
we use matmul function in Tensorflow
layer = tf.matmul(input, weights) + biases
if use_relu:
        layer = tf.nn.relu(layer)
return layer
# Create all the layers
layer conv1 = make generator model(input=x,
num_input_channels=num_channels,
conv_filter_size=filter_size_conv1,
num_filters=num_filters_conv1)
```

```
layer_conv2 = make_generator_model(input=layer_conv1,
num input channels=num filters conv1,
conv_filter_size=filter_size_conv2,
num_filters=num_filters_conv2)
layer_conv3 = make_generator_model(input=layer_conv2,
num_input_channels=num_filters_conv2,
conv filter size=filter size conv3,
num filters=num filters conv3)
layer flat = create flatten layer(layer conv3)
layer fc1 = create fc layer(input=layer flat,
num inputs=layer flat.get shape()[1:4].num elements(),
num_outputs=fc_layer_size,
use relu=True)
layer_fc2 = create_fc_layer(input=layer_fc1,
num inputs=fc layer size,
num outputs=num_classes,
use relu=False)
y_pred = tf.nn.softmax(layer_fc2, name='y_pred')
y_pred_cls = tf.argmax(y_pred, dimension=1)
session.run(tf.compat.v1.global variables initializer())
cross entropy = tf.nn.softmax cross entropy with logits v2(logits=layer fc2,
labels=y_true)
cost = tf.reduce mean(cross entropy)
optimizer = tf.compat.v1.train.AdamOptimizer(learning rate=1e-4).minimize(cost)
correct prediction = tf.equal(y pred cls, y true cls)
accuracy = tf.reduce mean(tf.cast(correct prediction, tf.float32))
session.run(tf.compat.v1.global_variables_initializer())
# Display all stats for every epoch
def show_progress(epoch, feed_dict_train, feed_dict_validate, val loss,
total_epochs):
    acc = session.run(accuracy, feed_dict=feed_dict_train)
    val_acc = session.run(accuracy, feed_dict=feed_dict_validate)
    msg = "Training Epoch {0}/{4} --- Training Accuracy: {1:>6.1%}, Validation
Accuracy: {2:>6.1%}, Validation Loss: {3:.3f}"
print(msg.format(epoch + 1, acc, val_acc, val_loss, total_epochs))
total_iterations = 0
saver = tf.compat.v1.train.Saver()
print("")
# Training Function
def train(num iteration):
global total iterations
for i in range(total_iterations,
                   total_iterations + num_iteration):
```

```
x_batch, y_true_batch, _, cls_batch = data.train.next_batch(batch_size)
        x_valid_batch, y_valid_batch, _, valid_cls_batch =
data.valid.next_batch(batch_size)
        feed_dict_tr = {x: x_batch,
                        y_true: y_true_batch}
        feed_dict_val = {x: x_valid_batch,
                         y_true: y_valid_batch}
        session.run(optimizer, feed dict=feed dict tr)
if i % int(data.train.num examples / batch size) == 0:
            val loss = session.run(cost, feed dict=feed dict val)
            epoch = int(i / int(data.train.num_examples / batch_size))
# print(data.train.num_examples)
            # print(batch_size)
            # print(int(data.train.num_examples/batch_size))
            # print(i)
total_epochs = int(num_iteration / int(data.train.num_examples / batch_size)) + 1
show_progress(epoch, feed_dict_tr, feed_dict_val, val_loss, total_epochs)
            saver.save(session, 'trained_model')
    total iterations += num iteration
train(num_iteration=final_iter)
#except Exception as e:
    #print("Exception:",e)
# Calculate execution time
end = time.time()
dur = end-start
print("")
if dur<60:
print("Execution Time:",dur,"seconds")
elif dur>60 and dur<3600:
    dur=dur/60
print("Execution Time:",dur,"minutes")
    dur=dur/(60*60)
print("Execution Time:",dur,"hours")
from flask import Flask, render_template, flash, request, session,send_file
from flask import render_template, redirect, url_for, request
import warnings
import datetime
import cv2
import tensorflow as tf
import numpy as np
from tkinter import *
import os
app = Flask(__name__
app.config['DEBUG']
```

```
app.config['SECRET_KEY'] = '7d441f27d441f27567d441f2b6176a'
@app.route("/")
def homepage():
return render_template('index.html')
@app.route("/Test")
def Test():
return render_template('Test.html')
@app.route("/train", methods=['GET', 'POST'])
def train():
if request.method == 'POST':
import model as model
return render template('Tranning.html')
@app.route("/testimage", methods=['GET', 'POST'])
def testimage():
if request.method == 'POST':
        file = request.files['fileupload']
        file.save('data/alien_test/Test.jpg')
img = cv2.imread('data/alien_test/Test.jpg')
        train path = r'data\train'
if not os.path.exists(train_path):
print("No such directory")
raise Exception
# Path of testing images
dir_path = r'data\alien_test'
if not os.path.exists(dir_path):
print("No such directory")
raise Exception
# Walk though all testing images one by one
for root, dirs, files in os.walk(dir path):
for name in files:
print("")
                image_path = name
                filename = dir_path + '\\' + image_path
```

```
print(filename)
               image_size = 128
num channels = 3
images = []
if os.path.exists(filename):
# Reading the image using OpenCV
image1 = cv2.imread(filename)
                   import file path = filename
                   image = cv2.imread(import file path)
                   fnm = os.path.basename(import file path)
                   filename = 'Test.jpg'
cv2.imwrite(filename, image)
# print("After saving image:")
img = cv2.imread(import_file_path)
if img is None:
print('no data')
                   img1 = cv2.imread(import file path)
print(img.shape)
                   img = cv2.resize(img, ((int)(img.shape[1] / 5),
(int)(img.shape[0] / 5)))
                   original = img.copy()
                   neworiginal = img.copy()
                   cv2.imshow('original', img1)
                   gray = cv2.cvtColor(img1, cv2.COLOR_BGR2GRAY)
                   cv2.imshow('Original image', img1)
                   orimage = 'static/Out/Test.jpg'
cv2.imwrite(orimage, img1)
                   cv2.imshow('Gray image', gray)
                   gry = 'static/Out/gry.jpg'
cv2.imwrite(gry, gray)
                   p = 0
for i in range(img.shape[0]):
for j in range(img.shape[1]):
                          B = img[i][j][0]
                          G = img[i][j][1]
                          R = img[i][j][2]
if (B >110 and G >110 and R >110):
totalpixels = img.shape[0] * img.shape[1]
                   per_white = 100 * p / totalpixels
if per_white >10:
                       img[i][j] = [500, 300, 200]
                       cv2.imshow('color change', img)
```

```
# Guassian blur
blur1 = cv2.GaussianBlur(img, (3, 3), 1)
# mean-shift algo
newimg = np.zeros((img.shape[0], img.shape[1], 3), np.uint8)
                    criteria = (cv2.TERM_CRITERIA_EPS +
cv2.TERM_CRITERIA_MAX_ITER, 10, 1.0)
                    img = cv2.pyrMeanShiftFiltering(blur1, 20, 30, newimg, 0,
criteria)
                    cv2.imshow('means shift image', img)
                    noise = 'static/Out/noise.jpg'
cv2.imwrite(noise, img)
# Guassian blur
blur = cv2.GaussianBlur(img, (11, 11), 1)
                    blur = cv2.GaussianBlur(img, (11, 11), 1)
# Canny-edge detection
canny = cv2.Canny(blur, 160, 290)
                    canny = cv2.cvtColor(canny, cv2.COLOR_GRAY2BGR)
# contour to find leafs
bordered = cv2.cvtColor(canny, cv2.COLOR_BGR2GRAY)
                    contours, hierarchy = cv2.findContours(bordered,
cv2.RETR TREE, cv2.CHAIN APPROX NONE)
                    maxC = 0
for x in range(len(contours)):
if len(contours[x]) > maxC:
                            maxC = len(contours[x])
                            maxid = x
perimeter = cv2.arcLength(contours[maxid], True)
# print perimeter
Tarea = cv2.contourArea(contours[maxid])
                    cv2.drawContours(neworiginal, contours[maxid], -1, (0, 0,
255))
                    cv2.imshow('Contour', neworiginal)
# cv2.imwrite('Contour complete leaf.jpg',neworiginal)
                    # Creating rectangular roi around contour
height, width, _ = canny.shape
                    min_x, min_y = width, height
                    \max x = \max y = 0
frame = canny.copy()
# computes the bounding box for the contour, and draws it on the frame,
for contour, hier in zip(contours, hierarchy):
                        (x, y, w, h) = cv2.boundingRect(contours[maxid])
                        min x, max_x = min(x, min_x), max(x + w, max_x)
                        min_y, max_y = min(y, min_y), max(y + h, max_y)
if w >80 and h >80:
# cv2.rectangle(frame, (x,y), (x+w,y+h), (255, 0, 0), 2) #we do not draw the
rectangle as it interferes with contour later on
roi = img[y:y + h, x:x + w]
                            originalroi = original[y:y + h, x:x + w]
if (\max x - \min x > 0 \text{ and } \max y - \min y > 0):
                        roi = img[min_y:max_y, min_x:max_x]
                        originalroi = original[min_y:max_y, min_x:max_x]
                        cv2.rectangle(frame, (min_x, min_y), (max_x, max_y), (255,
0, 0),
2) # we do not draw the rectangle as it interferes with contour
```

```
cv2.imshow('ROI', frame)
                    roi12 = 'static/Out/roi.jpg'
cv2.imwrite(roi12, frame)
                    cv2.imshow('rectangle ROI', roi)
img = roi
# Changing colour-space
                    # imghsv = cv2.cvtColor(img, cv2.COLOR_BGR2HSV)
imghls = cv2.cvtColor(roi, cv2.COLOR BGR2HLS)
                    cv2.imshow('HLS', imghls)
                    imghls[np.where((imghls == [30, 200, 2]).all(axis=2))] = [0,
200, 0]
                    cv2.imshow('new HLS', imghls)
# Only hue channel
huehls = imghls[:, :, 0]
                    cv2.imshow('img_hue hls', huehls)
# ret, huehls = cv2.threshold(huehls,2,255,cv2.THRESH_BINARY)
huehls[np.where(huehls == [0])] = [35]
                    cv2.imshow('img_hue with my mask', huehls)
# Thresholding on hue image
ret, thresh = cv2.threshold(huehls, 28, 255, cv2.THRESH_BINARY_INV)
                    cv2.imshow('thresh', thresh)
# Masking thresholded image from original image
mask = cv2.bitwise_and(originalroi, originalroi, mask=thresh)
                    cv2.imshow('masked out img', mask)
# Resizing the image to our desired size and preprocessing will be done exactly as
done during training
image = cv2.resize(image1, (image_size, image_size), 0, 0, cv2.INTER_LINEAR)
                    images.append(image)
                    images = np.array(images, dtype=np.uint8)
                    images = images.astype('float32')
                    images = np.multiply(images, 1.0 / 255.0)
# The input to the network is of shape [None image_size image_size num_channels].
Hence we reshape.
x_batch = images.reshape(1, image_size, image_size, num_channels)
# Let us restore the saved model
sess = tf.compat.v1.Session()
# Step-1: Recreate the network graph. At this step only graph is created.
saver = tf.compat.v1.train.import_meta_graph('models/trained_model.meta')
# Step-2: Now let's load the weights saved using the restore method.
saver.restore(sess, tf.train.latest_checkpoint('./models/'))
# Accessing the default graph which we have restored
graph = tf.compat.v1.get_default_graph()
# Now, let's get hold of the op that we can be processed to get the output.
                    # In the original network y pred is the tensor that is the
prediction of the network
y_pred = graph.get_tensor_by_name("y_pred:0")
## Let's feed the images to the input placeholders
x = graph.get_tensor_by_name("x:0")
```

```
y_true = graph.get_tensor_by_name("y_true:0")
                    y_test_images = np.zeros((1, len(os.listdir(train_path))))
# Creating the feed_dict that is required to be fed to calculate y_pred
feed_dict_testing = {x: x_batch, y_true: y_test_images}
                    result = sess.run(y_pred, feed_dict=feed_dict_testing)
# Result is of this format [[probabiliy_of_classA probability_of_classB ....]]
print(result)
# Convert np.array to list
a = result[0].tolist()
                    r = 0
# Finding the maximum of all outputs
max1 = max(a)
                    index1 = a.index(max1)
                    predicted_class = None
# Walk through directory to find the label of the predicted output
for root, dirs, files in os.walk(train_path):
for name in dirs:
if count == index1:
                                predicted class = name
                            count += 1
# If the maximum confidence output is largest of all by a big margin then
                    # print the class or else print a warning
for i in a:
if i != max1:
if max1 - i < i:
                                r = 1
out = ''
pre = ""
if r == 0:
print(predicted_class)
if (predicted class == "Black spot"):
                            out = predicted_class
                            pre = 'Griffin Fertilizer reducing the fungus'
elif (predicted_class == "canker"):
                            out = predicted_class
                            pre = 'sprayed with Bordeaux mixture 1.0 per cent.'
elif (predicted_class == "greening"):
                            out = predicted_class
                            pre = 'Mn-Zn-Fe-B micronutrient fertilizer'
elif (predicted class == "healthy"):
                            out = predicted class
# messagebox.showinfo("Uses", '')
elif (predicted_class == "Melanose"):
                            out = predicted_class
                            pre = 'strobilurin fungicide'
```

```
else:
                        out = 'Could not classify with definite confidence'
else:
print("File does not exist")
        org = 'static/Out/Test.jpg'
gry ='static/Out/gry.jpg'
noise = 'static/Out/noise.jpg'
roi12 = 'static/Out/roi.jpg'
return
render_template('Test.html',result=out,org=org,gry=gry,inv=noise,noi=roi12,fer=pre
def sendmsg(targetno,message):
import requests
requests.post("http://smsserver9.creativepoint.in/api.php?username=fantasy&passwor
d=596692&to=" + targetno + "&from=FSSMSS&message=Dear user your msg is " +
message + " Sent By FSMSG
FSSMSS&PEID=1501563800000030506&templateid=1507162882948811640")
if __name__ == '__main__':
    app.run(debug=True, use reloader=True)
import cv2
import os
import glob
from sklearn.utils import shuffle
import numpy as np
```

```
def load_train(train_path, image_size, classes):
    images = []
    labels = []
    img_names = []
    cls = []
print('Going to read training images')
for fields in classes:
        index = classes.index(fields)
print('Now going to read {} files (Index: {})'.format(fields, index))
        path = os.path.join(train_path, fields, '*g')
        files = glob.glob(path)
for fl in files:
            image = cv2.imread(f1)
            image = cv2.resize(image, (image_size, image_size),0,0,
cv2.INTER_LINEAR)
            image = image.astype(np.float32)
            image = np.multiply(image, 1.0 / 255.0)
            images.append(image)
            label = np.zeros(len(classes))
            label[index] = 1.0
labels.append(label)
            flbase = os.path.basename(fl)
            img names.append(flbase)
            cls.append(fields)
    images = np.array(images)
    labels = np.array(labels)
    img_names = np.array(img_names)
    cls = np.array(cls)
return images, labels, img_names, cls
class DataSet(object):
def __init__(self, images, labels, img_names, cls):
self._num_examples = images.shape[0]
self._images = images
self._labels = labels
self. img names = img names
self. cls = cls
self._epochs_done = 0
self._index_in_epoch = 0
@property
def images(self):
return self._images
@property
def labels(self):
return self. labels
@property
def img names(self):
return self._img_names
@property
```

```
def cls(self):
return self._cls
@property
def num_examples(self):
return self._num_examples
@property
def epochs done(self):
return self._epochs_done
def next_batch(self, batch_size):
"""Return the next `batch_size` examples from this data set."""
start = self._index_in_epoch
self._index_in_epoch += batch_size
if self._index_in_epoch >self._num_examples:
# After each epoch we update this
self. epochs done += 1
start = 0
self._index_in_epoch = batch_size
assert batch_size <= self._num_examples</pre>
    end = self._index_in_epoch
return self._images[start:end], self._labels[start:end],
self._img_names[start:end], self._cls[start:end]
def read_train_sets(train_path, image_size, classes, validation_size):
class DataSets(object):
pass
data sets = DataSets()
  images, labels, img_names, cls = load_train(train_path, image_size, classes)
  images, labels, img names, cls = shuffle(images, labels, img names, cls)
if isinstance(validation size, float):
    validation size = int(validation size * images.shape[0])
 validation images = images[:validation size]
  validation_labels = labels[:validation_size]
  validation img names = img names[:validation size]
 validation cls = cls[:validation size]
 train_images = images[validation_size:]
  train_labels = labels[validation_size:]
  train_img_names = img_names[validation_size:]
 train_cls = cls[validation_size:]
 data sets.train = DataSet(train images, train labels, train img names,
train cls)
  data_sets.valid = DataSet(validation_images, validation_labels,
validation img names, validation cls)
return data sets
import tensorflow as tf
import numpy as np
from tkinter import *
```

```
import os
from tkinter import filedialog
import cv2
import time
from matplotlib import pyplot as plt
from tkinter import messagebox
def endprogram():
print ("\nProgram terminated!")
   sys.exit()
def training():
import Training as tr
def imgtraining():
    import file path = filedialog.askopenfilename()
    image = cv2.imread(import file path)
    filename = 'Test.jpg'
cv2.imwrite(filename, image)
print("After saving image:")
    gray = cv2.cvtColor(image, cv2.COLOR BGR2GRAY)
    cv2.imshow('Original image', image)
    cv2.imshow('Gray image', gray)
# import_file_path = filedialog.askopenfilename()
print(import_file_path)
    fnm = os.path.basename(import file path)
print(os.path.basename(import_file_path))
from PIL import Image, ImageOps
    im = Image.open(import file path)
    im invert = ImageOps.invert(im)
    im invert.save('lena_invert.jpg', quality=95)
    im = Image.open(import_file_path).convert('RGB')
    im_invert = ImageOps.invert(im)
    im_invert.save('tt.png')
    image2 = cv2.imread('tt.png')
```

```
cv2.imshow("Invert", image2)
""""______"""
img = image
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
   cv2.imshow('Original image', img)
#cv2.imshow('Gray image', gray)
dst = cv2.fastNlMeansDenoisingColored(img, None, 10, 10, 7, 21)
   cv2.imshow("Nosie Removal", dst)
gray = cv2.cvtColor(img, cv2.COLOR BGR2GRAY)
img = cv2.imread(import_file_path)
if img is None:
print('no data')
   img1 = cv2.imread(import_file_path)
print(img.shape)
   img = cv2.resize(img, ((int)(img.shape[1] / 5), (int)(img.shape[0] / 5)))
original = img.copy()
neworiginal = img.copy()
   cv2.imshow('original', img1)
   gray = cv2.cvtColor(img1, cv2.COLOR_BGR2GRAY)
   cv2.imshow('Original image', img1)
# cv2.imshow('Gray image', gray)
p = 0
for i in range(img.shape[0]):
for j in range(img.shape[1]):
           B = img[i][i][0]
           G = img[i][j][1]
           R = img[i][j][2]
if (B >110 and G >110 and R >110):
              p += 1
totalpixels = img.shape[0] * img.shape[1]
   per_white = 100 * p / totalpixels
if per_white >10:
       img[i][j] = [500, 300, 200]
       cv2.imshow('color change', img)
# Guassian blur
blur1 = cv2.GaussianBlur(img, (3, 3), 1)
# mean-shift algo
newimg = np.zeros((img.shape[0], img.shape[1], 3), np.uint8)
   criteria = (cv2.TERM_CRITERIA_EPS + cv2.TERM_CRITERIA_MAX_ITER, 10, 1.0)
   img = cv2.pyrMeanShiftFiltering(blur1, 20, 30, newimg, 0, criteria)
   cv2.imshow('means shift image', img)
# Guassian blur
blur = cv2.GaussianBlur(img, (11, 11), 1)
   cv2.imshow('Noise Remove', blur)
   corners = cv2.goodFeaturesToTrack(gray, 27, 0.01, 10)
   corners = np.int0(corners)
```

```
# we iterate through each corner,
    # making a circle at each point that we think is a corner.
for i in corners:
        x, y = i.ravel()
        cv2.circle(image, (x, y), 3, 255, -1)
    plt.imshow(image), plt.show()
def testing():
global testing screen
    testing_screen = Toplevel(main_screen)
    testing_screen.title("Testing")
# Login_screen.geometry("400x300")
testing_screen.geometry("600x450+650+150")
    testing screen.minsize(120, 1)
    testing_screen.maxsize(1604, 881)
    testing screen.resizable(1, 1)
# login_screen.title("New Toplevel")
Label(testing screen, text='''Upload Image''', background="#d9d9d9",
disabledforeground="#a3a3a3",
foreground="#000000", bg="turquoise", width="300", height="2", font=("Calibri",
16)).pack()
    Label(testing_screen, text="").pack()
    Label(testing_screen, text="").pack()
    Label(testing screen, text="").pack()
    Button(testing_screen, text='''Upload Image''', font=(
'Verdana', 15), height="2", width="30", command=imgtest).pack()
def imgtest():
    import file path = filedialog.askopenfilename()
    image = cv2.imread(import file path)
print(import_file_path)
    filename = 'data/alien test/Test.jpg'
cv2.imwrite(filename, image)
print("After saving image:")
def main account screen():
from PIL import Image, ImageTk
global main screen
    main screen = Tk()
    width = 600
height = 600
screen_width = main_screen.winfo_screenwidth()
    screen_height = main_screen.winfo_screenheight()
```

```
x = (screen_width / 2) - (width / 2)
y = (screen_height / 2) - (height / 2)
main_screen.geometry("%dx%d+%d+%d" % (width, height, x, y))
    main_screen.resizable(0, 0)
# main_screen.geometry("300x250")
main_screen.title("Leaf Disease classification")
    Label(text="Leaf Disease classification", bg="turquoise", width="300",
height="5", font=("Calibri", 16)).pack()
    Label(text="").pack()
    Label(text="").pack()
    image = ImageTk.PhotoImage(Image.open('gui/12344.jpg'))
    Label(main_screen, text='Hello', image=image, compound='left', height="100",
width="200",).pack()
    Button(text="Training", font=(
'Verdana', 15), height="2", width="30", command=training,
highlightcolor="black").pack(side=TOP)
    Label(text="").pack()
    Button(text="Testing", font=(
'Verdana', 15), height="2", width="30", command=testing).pack(side=TOP)
    Label(text="").pack()
    main screen.mainloop()
main account screen()
```

GitHub & Project Demo link:

GitHub:

IBM-EPBL/IBM-Project-12474-1659451958

Project Demo link:

https://www.youtube.com/watch?v=CMWIfIJVYyI

OUTPUT:

