

PROJECT REPORT

Fertilizers Recommendation for Disease Prediction

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1.INTRODUCTION :

- Agriculture is the most important sector in today's life. Most plants are affected by a wide variety of bacterial and fungal diseases. Diseases on plants placed a major constraint on the production and a major threat to food security. Hence, early and accurate identification of plant diseases is essential to ensure high quantity and best quality. In recent years, the number of diseases on plants and the degree of harm caused has increased due to the variation in pathogen varieties, changes in cultivation methods, and inadequate plant protection techniques.

1.1 Project Overview

- An Automated system is introduced to identify different diseases on plants by checking the symptoms shown on the leaves of the plant. Deep learning techniques are used to identify the diseases and suggest the precautions that can be taken for those diseases changes in cultivation method and inadequate plant protection techniques and suggest all the precautions that can be taken for those diseases.

1.2 Purpose

- To Detect and recognize the plant diseases and to recommend fertilizer, it is necessary to identify the diseases and to recommend to get different and useful features needed for the purpose of analyzing later.
- To provide symptoms in identifying the disease at its earliest. Hence the authors proposed and implemented new fertilizers Recommendation System for Crop Disease Prediction.

2.LITREATURE SURVEY

2.1 Existing Problem

- Adequate mineral nutrition is central to crop production. However, it can also exert considerable Influence on disease development. Fertilizer application can increase or decrease development of diseases caused by different pathogens, and the mechanisms responsible are complex, including effects of nutrients on plant growth, plant resistance mechanisms and direct effects on the pathogen. The effects of mineral nutrition on plant disease and the mechanisms responsible for those effects have been dealt with comprehensively elsewhere. In India, around 40% of land is kept and grown using reliable irrigation technologies, while the rest relies on the monsoon environment for water. Irrigation decreases reliance on the monsoon, increases food security, and boosts agricultural production.
- Most research articles use humidity, moisture, and temperature sensors near the plant's root, with an external device handling all of the data provided by the sensors and transmitting it directly to an Android application. It was created to measure the approximate values of temperature, humidity and moisture sensors that were programmed into a microcontroller to manage the amount of water.

2.2 References :

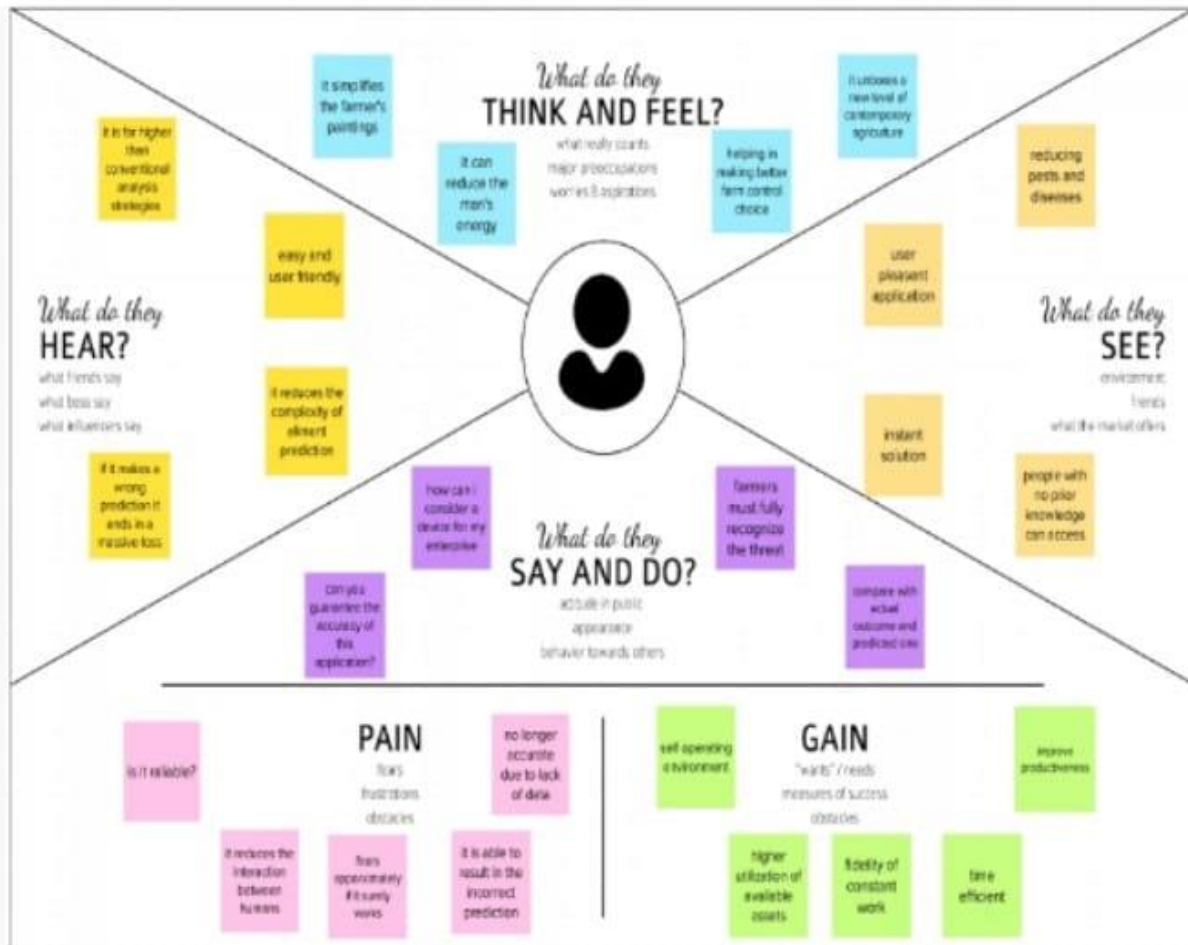
- Reyes Angie .K, Juan C. Caicedo, and Jorge E. Camargo, "Fine-tuning Deep Convolutional Networks for Plant Recognition", In CLEF (Working Notes), 2015.
- Hamrouni .L, Aiadi .O, Khaldi .B and Kherfi .M.L, "Plants Species Identification using Computer Vision Techniques", Revue des Bioressources 7, no. 1, 2018.
- 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 :

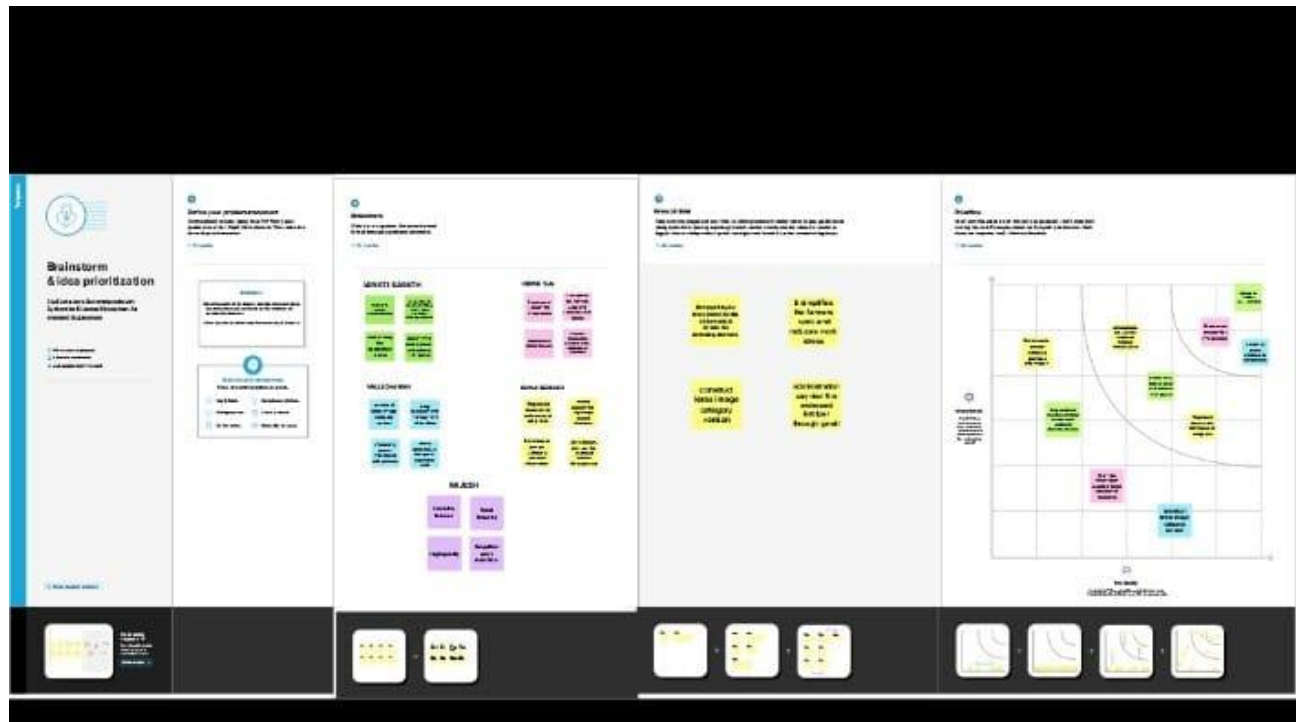
- The solution to the problem is Machine learning, which is one of the applications of Artificial Intelligence, is being used to implement the proposed system. Crop recommendation is going to recommend you the best crop you can grow in your land as per the soil nutrition value and along with as per the climate in that region. And recommending the best fertilizer for every particular crop is also a challenging task. And the other and most important issue is when a plant gets caught by heterogeneous diseases that effect on less amount of agriculture production and compromises with quality as well. To overcome all these issues this recommendation has been proposed .
- Nowadays a lot of research and work is being implemented in the smart and modern agriculture domain. Crop recommendation is characterized by a soil database comprised of Nitrogen, Phosphorus, potassium. The ensembles technique is used to build a recommendation model that combines the prediction of multiple machine learning. Models to recommend the right crop based on soil value and the best fertilizer to use.

3.IDEATION & PROPOSED SOLUTION :

3.1 Empathy Map Canvas :



3.2 Ideation & Brainstorming :



3.3 Proposed Solution :

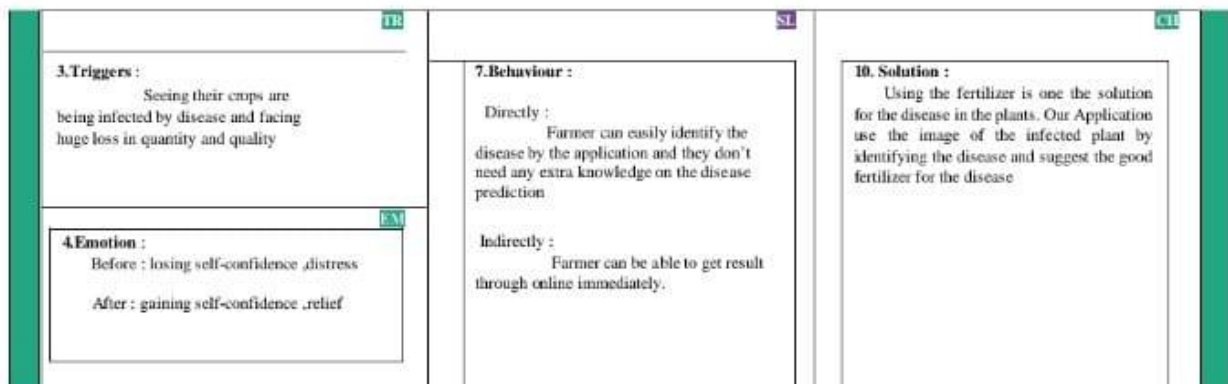
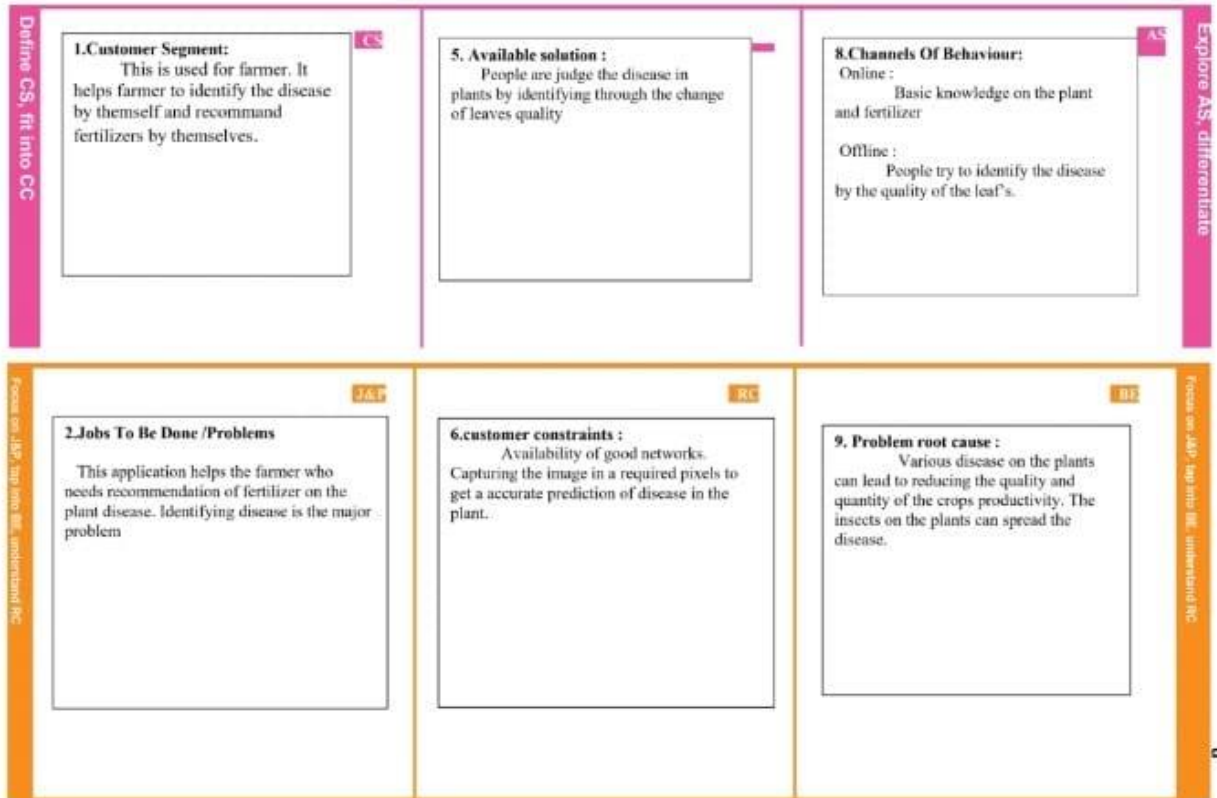
- The idea of the proposed solution uses Deep learning and Machine algorithm to classify leaves and identify the diseases and suggest the fertilizers. The deep learning process includes the MobileNetV2 and VGG19 training Models.
- Based on the leaf disease detected , the model recommendation for fertilizers for the prevention. The farmers and researchers are the end users get benefited by the system.
- More accurate in others. The system is more robust incorporating more image data sets with wider variations. This system also estimates the probability of infected plant.
- Plant growth can be enhanced. Ensure plants are getting supplied with every nutrient they need also and multiple cross in grow in every yields for every season. It also helps people's nutritional needs.

3.4 Problem Solution Fit

Project Title: Fertilizer Recommendation System For Disease Prediction

Project Design Phase-I - Solution Fit

Team ID: PNT2022TMID24348



4.REQUIREMENT ANALYSIS :

4.1.Functional Requirements

Following are the functional requirements of the proposed solution .

| Fr.no | Functional requirement | Sub requirement (story/subtask) |
|-------|------------------------|--|
| Fr-1 | User registration | Registration through form Registration through Gmail |
| Fr-2 | User confirmation | Confirmation via OTP Confirmation via Email |
| Fr-3 | Capturing image | Capture the image of the leaf And check the parameter of the captured image . |
| Fr-4 | Image processing | Upload the image for the prediction of the disease in the leaf. |
| Fr-5 | Leaf identification | Identify the leaf and predict the disease in leaf. |
| Fr-6 | Image description | Suggesting the best fertilizer for the disease . |

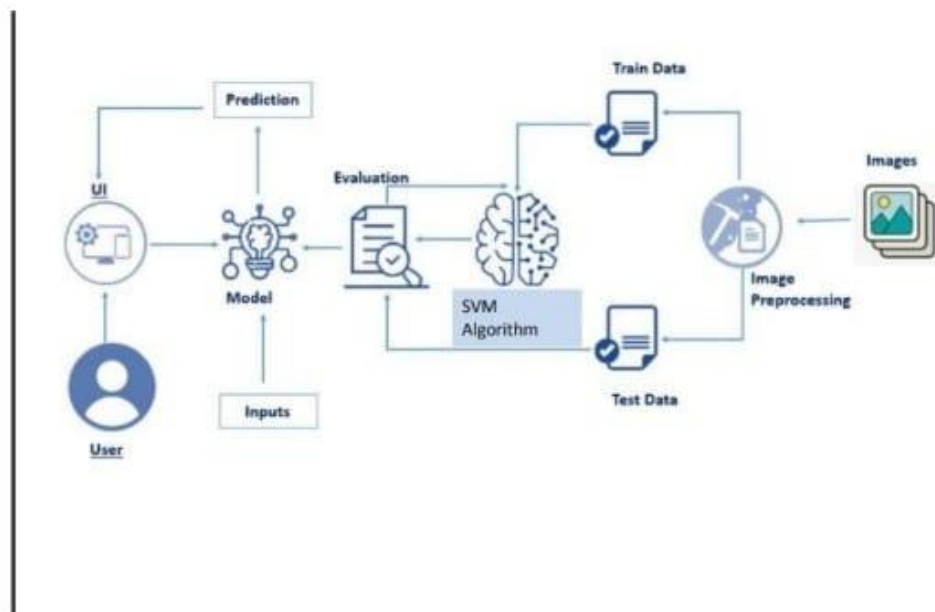
4.2 Non Functional Requirements

Following are the non-functional requirement of the proposed solution

| NFr.no | Non-functional requirement | Description |
|--------|----------------------------|---|
| Nfr-1 | Usability | Datasets of all the leaf is used to detecting the disease that present in the leaf. |
| Nfr-2 | Security | The information belongs to the user and leaf are secured highly. |
| Nfr-3 | Reliability | The leaf quality is important for the predicting the disease in leaf. |
| Nfr-4 | Performance | The performance is based on the quality of the leaf used for disease prediction |
| Nfr-5 | Availability | It is available for all user to predict the disease in the plant |
| Nfr-6 | Scalability | Increasing the prediction of the disease in the leaf |

5.PROJECT DESIGN :

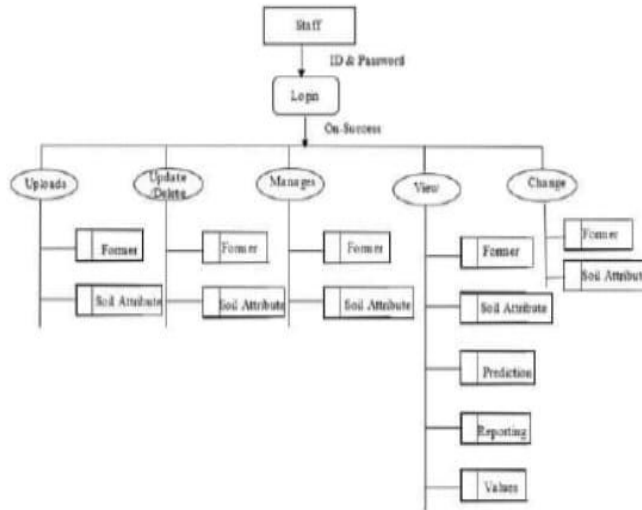
5.1 Solution & Technical Architecture



5.2 Data Flow Daigrams

A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.

Example:



5.3 User Stories

Use the below template to list all the user stories for the product.

| User Type | Functional Requirement (Epic) | User Story Number | User Story / Task | Acceptance criteria | Priority | Release |
|------------------------|-------------------------------|-------------------|--|--|----------|----------|
| Customer (Mobile user) | Registration | USN-1 | As a user, I can register for the application by entering my email, password, and confirming my password. | I can access my account / dashboard | High | Sprint-1 |
| | | USN-2 | As a user, I will receive confirmation email once I have registered for the application | I can receive confirmation email & click confirm | High | Sprint-1 |
| | | USN-3 | As a user, I can register for the application through Google | I can register & access the dashboard with Google Login | Low | Sprint-2 |
| | | USN-4 | As a user, I can register for the application through Gmail | | Medium | Sprint-1 |
| | Login | USN-5 | As a user, I can log into the application by entering email & password | | High | Sprint-1 |
| Customer (Web user) | Dashboard | | | | | |
| | | USN-6 | As a User can view the dash board ,and this dashboard include the check roles of access and then move to the manage modules. | I can view the dashboard in this fertilizer recommendation system for disease prediction | Medium | |
| | | USN-7 | | | | |
| Administrator | | | As a user once view the manage modules this describes the Manage system Admins and Manage Roles of User and etc.. | | | |

6.PROJECT PLANNING & SCHEDULING :

6.1 Sprint Planning and Estimation

| Sprint | Functional Requirement (Epic) | User Story Number | User Story / Task | Story Points (Total) | Priority | 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 | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | Model Creation and Training (Vegetables) | | Create a model which can classify diseased vegetable plants from given images | 2 | High | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |

| Sprint | Functional Requirement (Epic) | User Story Number | User Story / Task | Story Points (Total) | Priority | 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 | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | Registration | USN-1 | As a user, I can register by entering my email, password, and confirming my password or via OAuth API | 3 | Medium | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | Upload page | USN-2 | As a user, I will be redirected to a page where I can upload my pictures of crops | 4 | High | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | Suggestion results | USN-3 | As a user, I can view the results and then obtain the suggestions provided by the ML model | 4 | High | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | Base Flask App | | A base Flask web app must be created as an interface for the ML model | 2 | High | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| Sprint-3 | Login | USN-4 | As a user/admin/shopkeeper, I can log into the application by entering email & password | 2 | High | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | User Dashboard | USN-5 | As a user, I can view the previous results and history | 3 | Medium | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |

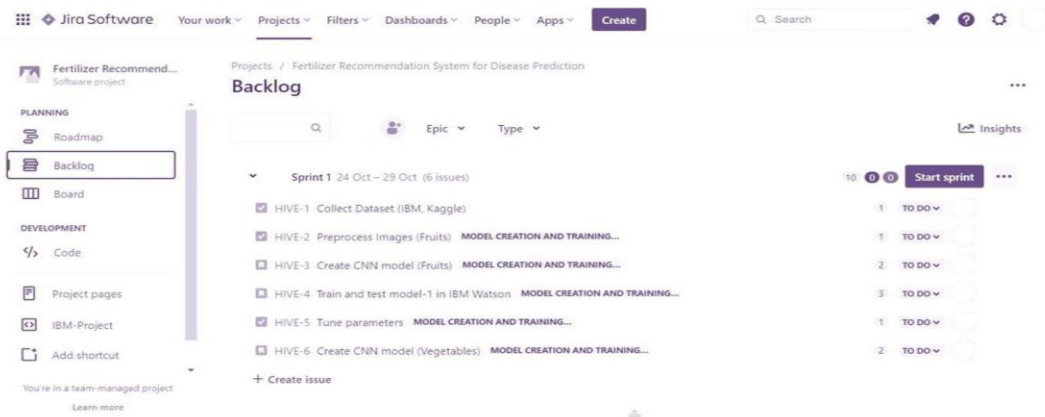
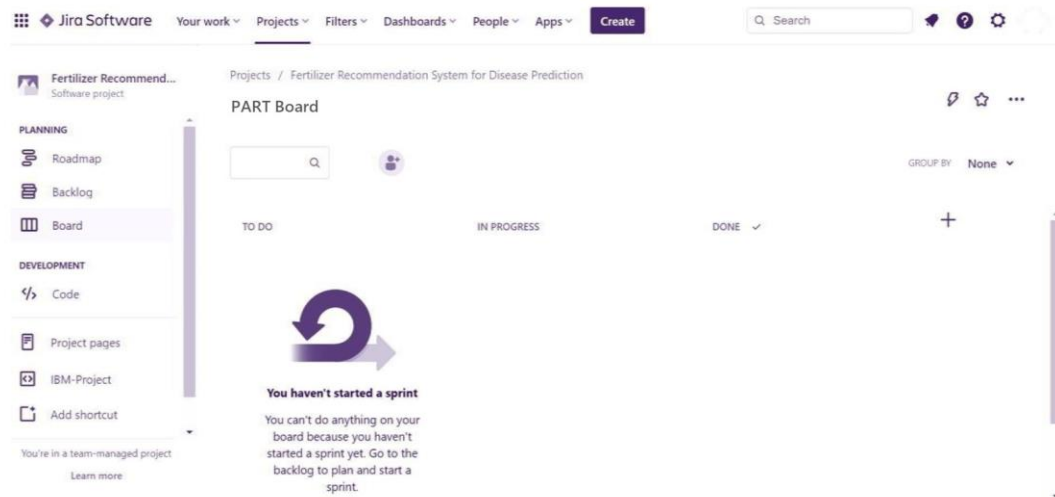
| | | | | | | |
|----------|------------------------|-------|---|---|--------|---|
| | Integration | | Integrate Flask, CNN model with Cloudant DB | 5 | Medium | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | Containerization | | Containerize Flask app using Docker | 2 | Low | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| Sprint-4 | Dashboard (Admin) | USN-6 | As an admin, I can view other user details and uploads for other purposes | 2 | Medium | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | Dashboard (Shopkeeper) | USN-7 | As a shopkeeper, I can enter fertilizer products and then update the details if any | 2 | Low | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | Containerization | | Create and deploy Helm charts using Docker Image made before | 2 | Low | Guna sekhar, Sarath, Charan, Rajesh, Hema sai |
| | | | | | | |

6.2 Sprint Delivery Schedule

| 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 |

NOTE: Burndown charts, Velocity to be updated dynamically after end of sprints Roadmap:

6.3 Reports from JIRA



7.CODING & SOLUTIONING:

7.1 Python – App.py :

```
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

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 = '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()
crop_recommendation_model_path = 'models/RandomForest.pkl'
crop_recommendation_model = pickle.load(
    open(crop_recommendation_model_path, 'rb'))
def weather_fetch(city_name):
    api_key = config.weather_api_key
    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)
        return temperature
    else:
        return None
def predict_image(img, model=disease_model):
    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

app=Flask(name)
@app.route('/crop-predict', methods=['POST'])
def crop_prediction():
    title = 'Harvestify - 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)

@app.route('/fertilizer-predict', methods=['POST'])
def fert_recommend():
    title = 'Harvestify - 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)

@app.route('/disease-predict', methods=['GET', 'POST'])
def upload():
    if request.method=='POST':
        f=request.files['image']
        basepath=os.path.dirname(file)
        filepath=os.path.join(basepath,'uploads',f.filename)
        f.save(filepath)
        print('File Save')
        img=image.load_img(filepath,target_size=(128,128))
        x=image.img_to_array(img)
        print('Image to gray')
        x=np.expand_dims(x,axis=0)
        plant=request.form['plant']
        if(plant=='vegetable'):
            model=load_model("vegetable.h5")

```

```

y=np.argmax(model.predict(x),axis=1)

df=pd.read_excel('precautions_veg.xlsx')

if(plant=='fruit'):

    model=load_model('fruit.h5')

    y=np.argmax(model.predict(x),axis=1)

    df=pd.read_excel('precautions_fruits.xlsx')

    return df.iloc[y[0]]['caution']

if name=='main':

    temp.run(debug=False)

```

7.2 Feature 1 :

Home.html:

```

<!DOCTYPE html>

<html lang="en">

<head>

    <meta charset="UTF-8">

    <meta http-equiv="X-UA-Compatible" content="IE=edge">

    <meta name="viewport" content="width=device-width, initial-scale=1.0">

    <title>home page</title>

    <style>

        body{

            margin: 0;

            padding: 0;

        }

        .container{

            padding: 30px 70px 30px 70px;

```

```
left: 20px;
right: 20px;
background-color: rgb(163, 192, 120);
font-size: 20pt;
font-family: 'Times New Roman';
```

```
}
```

```
.card{
    font: optional;
    display: flex;
```

```
W }
```

```
#h1{
    font-size: 50pt;
}
```

```
.menu{

    background-color: black;
```

```
}
```

```
#abc{
    color: white;
}
```

<p> Agriculture is one of the major sectors works wide.Over the years it has developed and the use of new technologies and equipment replaced almost all the traditional methods of farming.The plant diseases effect the production.Identification of diseases and taking necessary precautions is all done through naked eye,which requires labour and laboratries.This

application helps farmers in detecting the diseases by observing the spots on the leaves ,which inturn saves effort and labor costs.</p>

```

</div>
</div>
</div>
</body>
</html>
```

7.3 Feature 2:

Predict.html:

```
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <title>predict</title>
</head>

<style>
.container{
  display: flex;
  padding: 60px 70px 60px 70px;
}
.card{
  padding: 70px 80px 70px 80px;
```



```

}

.menu{
    padding: 10px 10px 10px 10px;
    background-color: black;
    color: white;
    font-size: 15pt;
}

</style>

<body>

    <div class="menu">

        <ul ><li>Plant disease Prediction</li></ul></div>

    <div class="container">

        <div class="card">

            <form>

                <h1>Drop in the image to get the Prediction </h1><br><br>

                <label><select name="Fruit" id="plant">

                    <option value="fruit" id="fruit">Fruit</option>

                    <option value="vagitable" id="vig">vegitable</option>

                </select>

                </label><br><br><br>

                <input id="default-btn" type="file" name=""
onchange="document.getElementById('output').src=window.URL.createObjectURL(this.files[0])
"><br><br><br>

                <img src="" id="output">

                <br>

                <button id="button" onclick ="display()" >Predict!</button><br><br>

```

</form>

</body>

</html>

8.TESTING

8.1 TEST CASES:

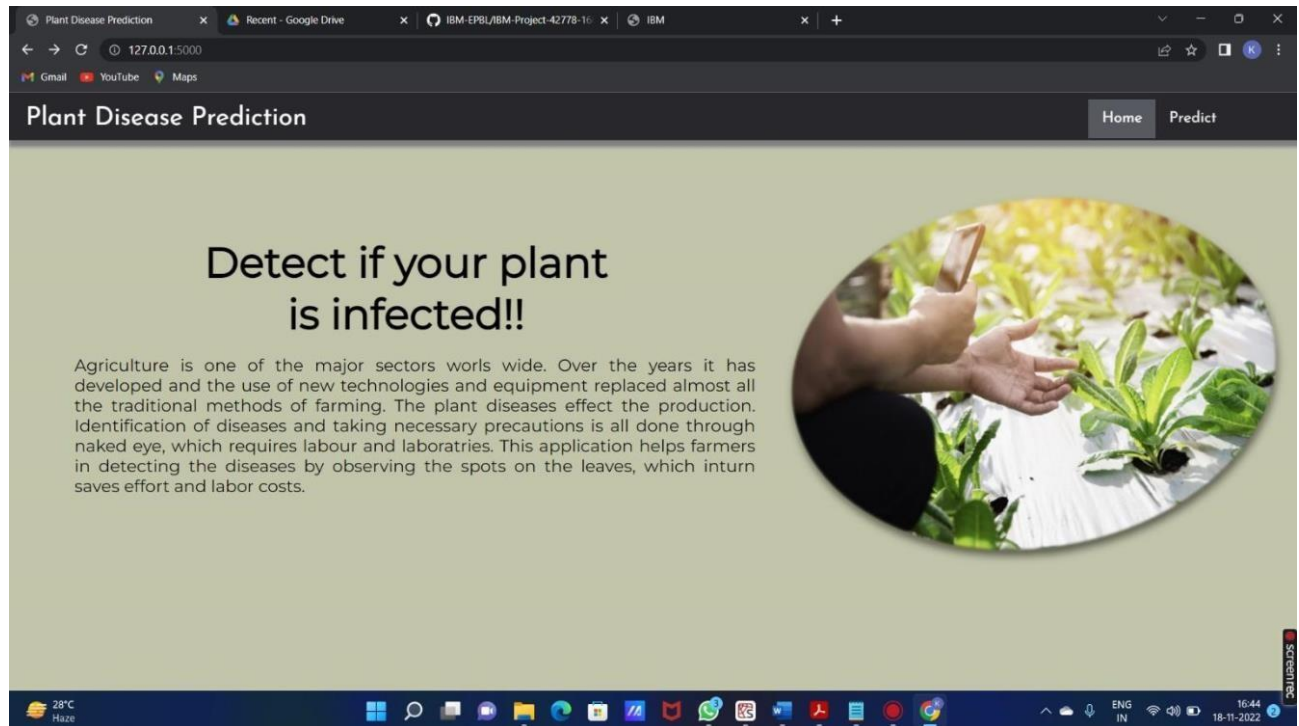
| SECTION | TOTAL CASES | NOT TESTED | FAIL | PASS |
|---------------------|-------------|------------|------|------|
| Leaf spots | 17 | 0 | 0 | 17 |
| Mosaic Leaf Pattern | 51 | 0 | 0 | 51 |
| Misshapen Leaves | 20 | 0 | 0 | 20 |
| Yellow Leaves | 7 | 0 | 0 | 7 |
| Fruit Rots | 9 | 0 | 0 | 9 |
| Fruit Spots | 4 | 0 | 0 | 4 |
| Blights | 2 | 0 | 0 | 2 |

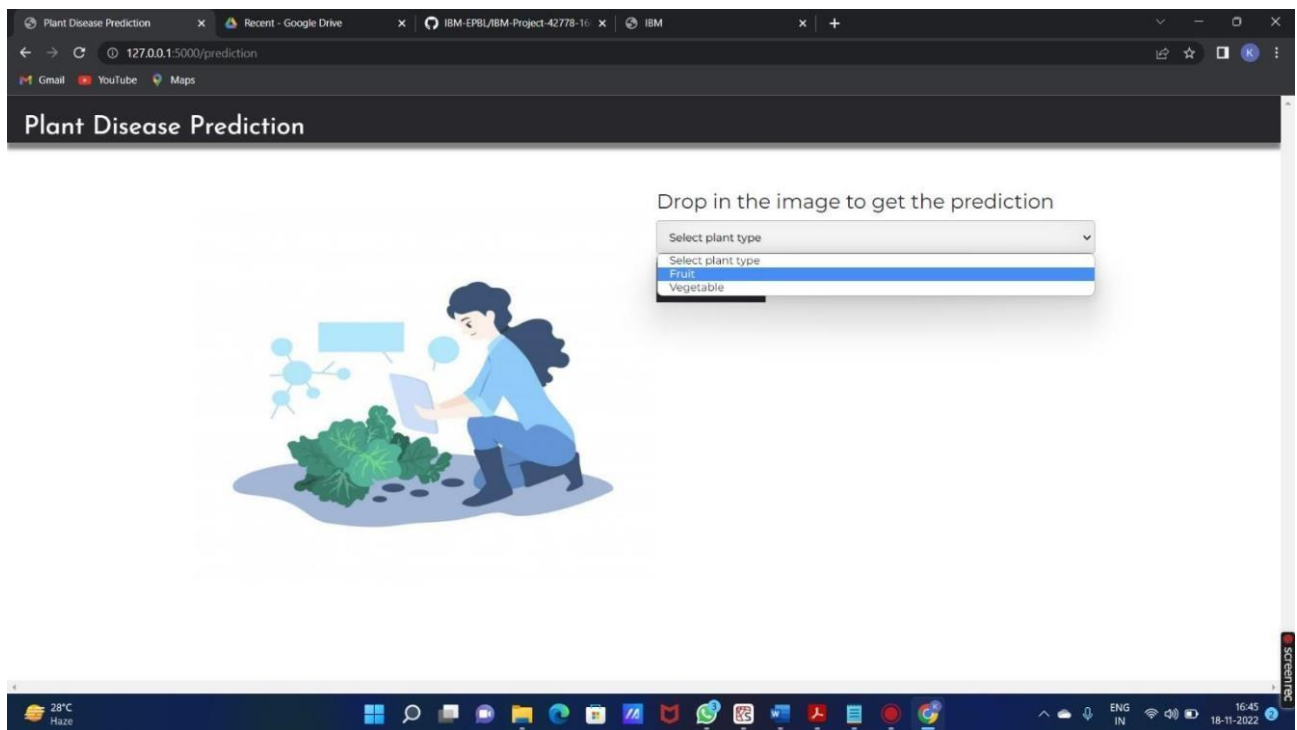
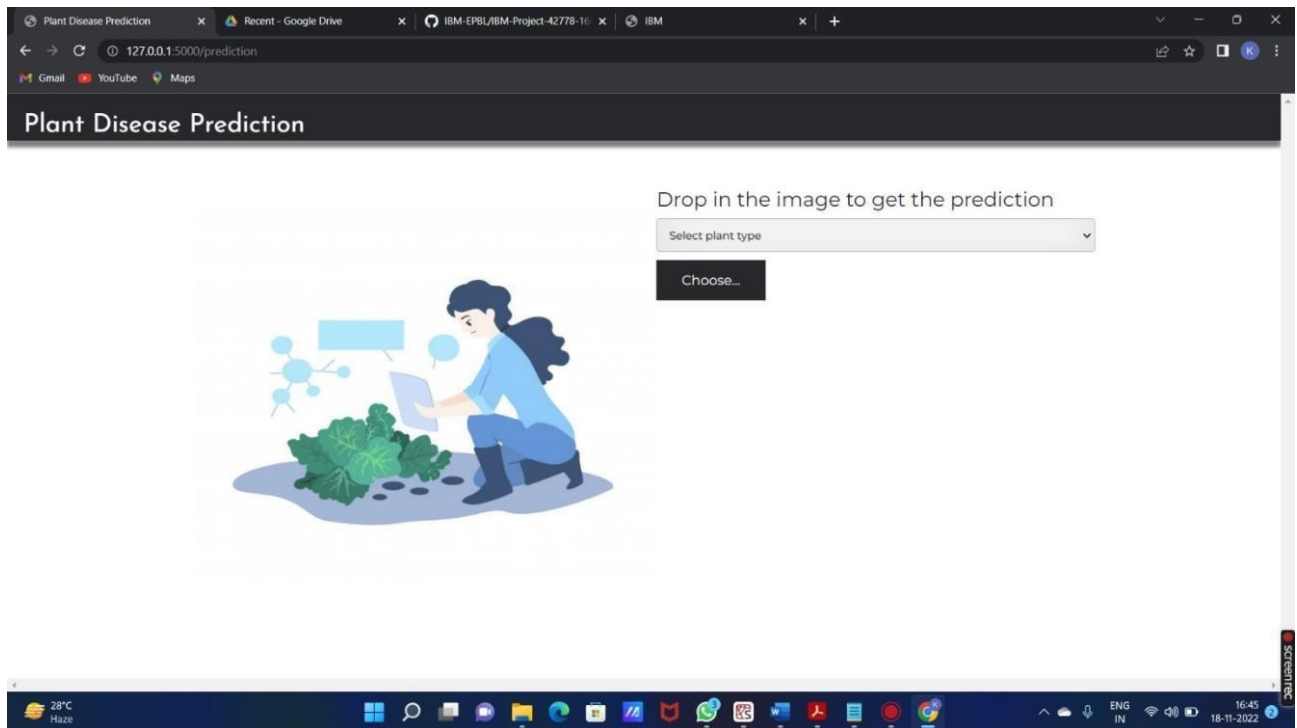
8.2 User Acceptance:

| RESOLUTION | SEVERITY 1 | SEVERITY 2 | SEVERITY 3 | SEVERITY 4 | SUBTOTAL |
|---------------------|------------|------------|------------|------------|----------|
| Leaf spots | 10 | 4 | 2 | 3 | 19 |
| Mosaic Leaf Pattern | 9 | 6 | 3 | 6 | 24 |
| 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 | 119 |

9.RESULTS

9.1 Performance Metrics







Plant Disease Prediction

Drop in the image to get the prediction

Fruit

Choose...



Prediction: Yaay!! Your apple plant is healthy. But, maintain the soil pH of 6.0 to 7.0 for healthy growth. Avoid planting apples in a low spot where cold air or frost can settle.

28°C Haze



16:45 18-11-2022

Plant Disease Prediction

Drop in the image to get the prediction

Vegetable

Choose...



Prediction: Ooops!! Your tomato plant is infected by Septoria leaf spot. Removing the infected leaves immediately will curb the spread of infection. Organic and chemical fungicides with chlorothalonil are effective in treatment.

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10.ADVANTAGES & DISADVANTAGES

List of advantages :

- The proposed model here produces very high accuracy of classification.
- Very large datasets can also be trained and tested.
- Images of very high can be resized within the proposed itself.

List of disadvantages :

- For training and testing, the proposed model requires very high computational time.
- The neural network architecture used in this project work has high complexity.

11.APPLICATIONS

1. The trained network model used to classify the image patterns with high accuracy.
2. The proposed model not only used for plant disease classification but also for other image pattern classification such as animal classification.
3. This project work application involves not only image classification but also for pattern recognition.

12.CONCLUSIONS

The model proposed here involves image classification of fruit datasets and vegetable datasets. The following points are observed during model testing and training:

- The accuracy of classification increased by increasing the number of epochs.
- For different batch sizes, different classification accuracies are obtained.
- The accuracies are increased by increasing more convolution layers.

- The accuracy of classification also increased by varying dense layers.
- Different accuracies are obtained by varying the size of kernel used in the convolution layer output.
- Accuracies are different while varying the size of the train and test datasets.

13.FUTURE SCOPE :

The proposed model in this project work can be extended to image recognition. The entire model can be converted to application software using python to exe software. The real time image classification, image recognition and video processing are possible with help OpenCV python library. This project work can be extended for security applications such as figure print recognition, iris recognition and face recognition.

14.APPENDIX

Source Code

(Jupyter notebook python code) fruit.ipynb (due to limited page size the code vegetable.ipynb uploaded in github)

```
#!/usr/bin/env python
# coding: utf-8
# In[1]: pwd
# In[2]: cd E:/IBM_MY_COURSE/Project/Dataset Plant Disease/fruitdataset/fruit-
dataset

# # Apply ImageDataGenerator functionality to Train and Test set
# # Preprocessing #

In[3]: from keras.preprocessing.image
import
```

```

ImageDataGenerator      train_datagen =
ImageDataGenerator(rescale=1./255,shear_range=0.2,zoom_range=0.2,horizontal_fli
p=True) test_datagen = ImageDataGenerator(rescale=1) # In[4]: pwd

# In[5]: x_train = train_datagen.flow_from_directory('E:/IBM_MY_COURSE/Project/Dataset
Plant Disease/fruit-
dataset/fruitdataset/train',target_size=(128,128),batch_size=32,class_mode='cate gorical')

# In[6]: x_test=test_datagen.flow_from_directory('E:/IBM_MY_COURSE/Project/Datas
et      Plant  Disease/fruit-dataset/fruit-dataset/test',target_size=(128,128),
batch_size=32,class_mode='categorical') # # Import the models

# In[7]:      from tensorflow.keras.models      import Sequential      from
tensorflow.keras.layers import Dense,Convolution2D,MaxPool2D,Flatten

# # Initializing the models 10
# In[8]: model=Sequential()
# # Add CNN Layers
# In[9]: model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu'))

# In[10]: x_train.class_indices
# # Add Pooling layer
# In[11]: model.add(MaxPool2D(pool_size=(2,2)))
# # Add Flatten layer # In[12]: model.add(Flatten())
# # Add Dense Layer
# In[21]:      model.add(Dense(40,      kernel_initializer='uniform',activation='relu'))
      model.add(Dense(20, kernel_initializer='random_uniform',activation='relu'))

# # Add Output Layer # In[24]:  model.add(Dense(6,activation='softmax',
      kernel_initializer='random_uniform'))

# # Compile the model
# In[25]: model.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accur acy'
])
# In[26]: len(x_train)

```



```

# In[27]: 5384/32
# # Fit the Model
# In[28]:
model.fit_generator(x_train,steps_per_epoch=168,validation_data=x_test,validation_steps=52,epochs=3)

# # Save the Model
# In[29]: model.save("fruit.h5")
# In[30]: ls
# # Test the Model
# In[32]: from keras.preprocessing import image from
tensorflow.keras.preprocessing.image import img_to_array from
tensorflow.keras.models import load_model import numpy as np

# In[33]: model = load_model("fruit.h5")
# # Test Apple_Healthy Class images
# In[37]: img = image.load_img('E:/IBM_MY_COURSE/Project/Dataset Plant
Disease/fruitdataset/fruit-dataset/test/Apple___healthy/00fca0da-2db3-
481bb98a9b67bb7b105c___RS_HL_7708.JPG',target_size=(128,128)) 11

# In[39]: x=image.img_to_array(img) x=np.expand_dims(x,axis=0)
# In[40]: pred = model.predict_classes(x)
# In[41]: pred
#
# In[45]: index
=['Apple___Black_rot','Apple___healthy','Corn_(maize)___Northern_Leaf_Blight','Corn_(
maize)___healthy','Peach___Bacterial_spot','Peach___healthy']

# In[46]: print('the given image belongs to=',index[pred[0]])
# # Test Apple Black Rot class images # In[54]: img =
image.load_img('E:/IBM_MY_COURSE/Project/Dataset Plant
Disease/fruitdataset/fruit-dataset/test/Apple___Black_rot/0f3d45f4-e121-42cda5b6-
be2f866a0574___JR_FrgE.S_2870.JPG',target_size=(128,128))

```

```

# In[55]: x=image.img_to_array(img) x=np.expand_dims(x,axis=0) pred =
model.predict_classes(x) print('the given image belongs to=',index[pred[0]])

# # Test Corn Northern leaf Blight class images
# In[56]: img = image.load_img('E:/IBM_MY_COURSE/Project/Dataset Plant
Disease/fruitdataset/fruitdataset/test/Corn_(maize)___Northern_Leaf_Blight/00a14441-
7a62-4034-bc40b196aeab2785___RS_NLB_3932.JPG',target_size=(128,128))

# In[57]: x=image.img_to_array(img) x=np.expand_dims(x,axis=0) pred =
model.predict_classes(x) print('the given image belongs to=',index[pred[0]])

# # Test Corn Healthy class images # In[58]: img =
image.load_img('E:/IBM_MY_COURSE/Project/Dataset Plant
Disease/fruitdataset/fruit-dataset/test/Corn_(maize)___healthy/0a68ef5a-027c41ae-b227-
159dae77d3dd___R.S_HL_7969 copy.jpg',target_size=(128,128))

# In[59]: x=image.img_to_array(img) x=np.expand_dims(x,axis=0) pred =
model.predict_classes(x) print('the given image belongs to=',index[pred[0]]) # #
Test Peach Bacterial spot class images # In[60]: img =
image.load_img('E:/IBM_MY_COURSE/Project/Dataset Plant
Disease/fruitdataset/fruit-dataset/test/Peach___Bacterial_spot/00ddc106-692e4c67-b2e8-
569c924caf49___Rutg._Bact.S_1228.JPG',target_size=(128,128)) 12 # In[61]:
x=image.img_to_array(img) x=np.expand_dims(x,axis=0) pred = model.predict_classes(x)
print('the given image belongs to=',index[pred[0]])

# # Test Peach Healthy class images
# In[62]: img = image.load_img('E:/IBM_MY_COURSE/Project/Dataset Plant
Disease/fruitdataset/fruit-dataset/test/Peach___healthy/1a07ce54-f4fd-41cfb088-
144f6bf71859___Rutg._HL_3543.JPG',target_size=(128,128))

# In[63]: x=image.img_to_array(img) x=np.expand_dims(x,axis=0) pred =
model.predict_classes(x) print('the given image belongs to=',index[pred[0]])

```

Github Link :- <https://github.com/IBM-EPBL/IBM-Project-54007-1661587406>

Demonstration Link:- <https://drive.google.com/file/d/13hJdeq0QtIT-NrE3QFPfR45rT-kwJgXq/view?usp=drivesdk>