



## CROP YIELD PREDICTION AND DISEASE DETECTION USING MACHINE LEARNING

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### Abstract

In this project we are combining plant leaf diseases image dataset with Crop Yield dataset to forecast or estimate future crop growth. To implement this project, we have used various crop attributes such as Potassium, Phosphorous and Nitrogen properties of soil and environment data such as Humidity, temperature, and PH values and then from crop leaf images we have extracted all and sum and average all infected leaf's and included in CROP dataset.

Above process Kernel Ridge dataset will be trained with various machine learning algorithms such as Lasso, Kernel Ridge, ENET and Stacking of all 3 algorithms and then evaluate each algorithm performance in terms of MEAN SQUARE ERROR (MSE) and in all algorithms ENET and LASSO giving less MSE values.

We are using OPENCV inbuilt API'S to cluster leaf data based on colour values and then these values will be weighted to get spots whose part is not in green colour. Below code is used to extract or identify infected parts from the leaf.

**Keywords:** Lasso, Kernel Ridge, ENET, OPENCV, MEAN SQUARE ERROR (MSE)

### 1. Introduction

Agriculture, since its invention and inception, is the prime and pre-eminent activity of every culture and civilization throughout the history of mankind. Not only is it a huge part of the expanding economy, but it is also necessary for our survival. It is also a vital sector for the future of humanity and the Indian economy. Additionally, it makes up a disproportionate amount of jobs. As time goes on, the amount of output that is needed has grown dramatically. People are misusing technology to make goods in large quantities. Every day, new kinds of hybrid types are created. However, these varieties do not provide the essential contents as naturally produced crop. These unnatural techniques spoil the soil. It all ends up in further environmental harm. Most of these unnatural techniques are wont to avoid losses. But when the producers of the crops know the accurate information on the crop yield it minimizes the loss.

Crop prediction is made by the system using historical data collection. The information is provided using historical data on weather, temperature, and several other variables. Our built application runs the algorithm and displays a list of crops with anticipated yield values that are appropriate for the entered data.

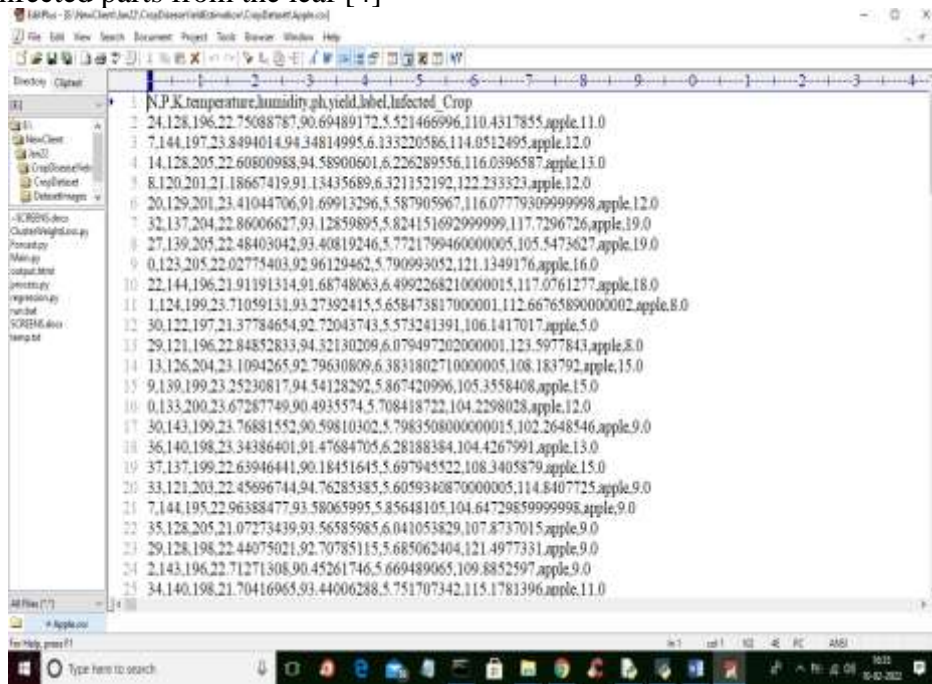
Using an Indian government dataset, Aruvansh Nigam, Saksham Garg, and Archit Agrawal [1] conducted trials and found that the Random Forest machine learning algorithm yields the best yield forecast accuracy. Simple Recurrent Neural Network, a sequential model, predicts rainfall more accurately than Long Short-Term Memory (LSTM) models. The study combines variables such as area, temperature, rainfall, and season to forecast yield. The best classifier, according to the results, is Random Forest when all the parameters are combined. Leo Brieman [2,3] is an expert in the random forest algorithm's strength, accuracy, and correlation.

### 2. Materials & Methods

#### 2.1 PROPOSED SYSTEM

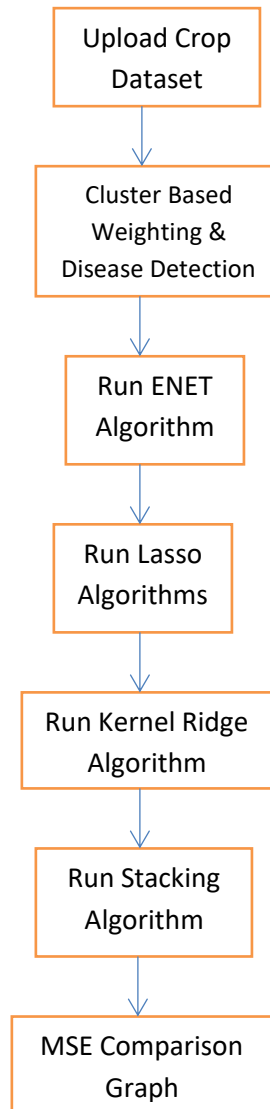
Above process dataset will be trained with various machine learning algorithms such as Lasso, Kernel Ridge, ENET and Stacking of all 3 algorithms and then evaluate each algorithm performance in terms of MEAN SQUARE ERROR (MSE) and in all algorithms ENET and LASSO giving less MSE values.

By using OPENCV inbuilt API'S to cluster leaf data based on colour values and then this value will be weighted to get spots whose part is not in green colour. Below code is used to extract or identify infected parts from the leaf [4]



```
1 N,P,K,temperature,humidity,ph,yield,label,infected_Crop
2 24.128,196.22,75088787,90.69489172,5.521466996,110.4317855,apple,11.0
3 7.144,197.23,8494014.94,34814995.6,133220586,114.0512495,apple,12.0
4 14.128,205.22,60800988,94.58900601,6.226289556,116.0396587,apple,13.0
5 8.120,201.21,18667419.91,13435689,6.321152192,122.233323,apple,12.0
6 20.129,201.21,41044706,91.69913296,5.587905967,116.07779309999998,apple,12.0
7 32.137,204.22,86006627,93.12859895,5.824151692999999,117.7296726,apple,19.0
8 27.139,205.22,48403042,93.40819246,5.7721799460000005,105.5473627,apple,19.0
9 0.123,205.22,02775403,92.96129462,5.790993052,121.1349176,apple,16.0
10 22.144,196.21,91191314,91.68748063,6.4992268210000015,117.0761277,apple,18.0
11 1.124,199.23,71059131,93.27392415,5.658473817000001,112.66765890000002,apple,8.0
12 30.122,197.21,37784654,92.72043743,5.573241391,106.1417017,apple,5.0
13 29.121,196.22,84852833,94.32130209,6.079497202000001,123.5977843,apple,8.0
14 13.126,204.23,1094265,92.79630809,6.3831802710000005,108.183792,apple,15.0
15 9.139,199.23,25230817,94.54128292,5.867420996,105.3558408,apple,15.0
16 0.133,200.23,67287749,90.4935574,5.708418722,104.2298028,apple,12.0
17 30.143,199.23,76881552,90.59810302,5.7983508000000015,102.2648546,apple,9.0
18 36.140,198.23,34386401,91.47684705,6.28188384,104.4267991,apple,13.0
19 37.137,199.22,63946441,90.18451645,5.697945522,108.3405879,apple,15.0
20 33.121,203.22,45696744,94.76285385,5.6059340870000005,114.8407725,apple,9.0
21 7.144,195.22,96388477,93.58065995,5.85648105,104.64729859999998,apple,9.0
22 35.128,205.21,07273439,93.56585985,6.041053829,107.8737015,apple,9.0
23 29.128,198.22,44075021,92.70785115,5.685062404,121.4977331,apple,9.0
24 2.143,196.22,71271308,90.45261746,5.669489065,109.8852597,apple,9.0
25 34.140,198.21,70416965,93.44006288,5.751707342,115.1781396,apple,11.0
```

In above screen first row represents dataset column names and remaining are the dataset values and in above screen N refers to Nitrogen, K refers to potassium and P refers to phosphorus soil properties and remaining values are the normal environment data with infected and yield crop[5].



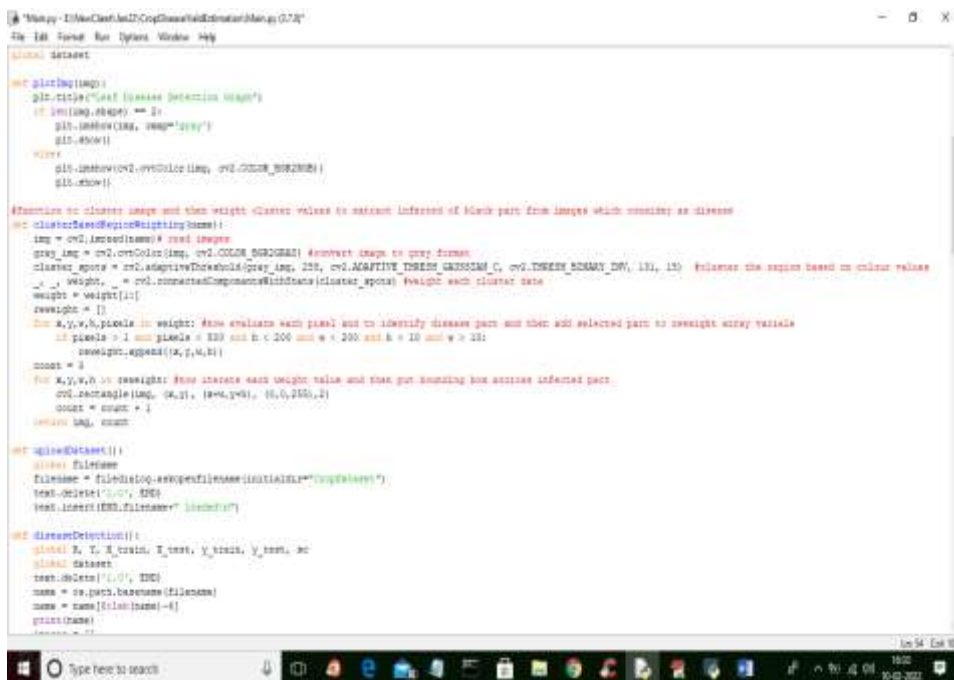
**Fig.3.1 Block diagram of Proposed Method**

### 3. Results and Discussion

In this present work we are combining plant leaf diseases image dataset with Crop Yield dataset to forecast or estimate future crop growth. To implement this project, we have used various crop attributes such as Potassium, Phosphorous and Nitrogen properties of soil and environment data such as Humidity, temperature and PH values and then from crop leaf images we have extracted all and sum and average all infected leaf's and included in CROP dataset[6].

Above process dataset will be trained with various machine learning algorithms such as Lasso, Kernel Ridge, ENET and Stacking of all 3 algorithms and then evaluate each algorithm performance in terms of MEAN SQUARE ERROR (MSE) and in all algorithms ENET and LASSO giving less MSE values.

By using OPENCV inbuilt API'S to cluster leaf data based on colour values and then these values will be weighted to get spots whose part is not in green colour. Below code is used to extract or identify infected parts from the leaf[7].



```
%% Load dataset
load('dataset.mat');

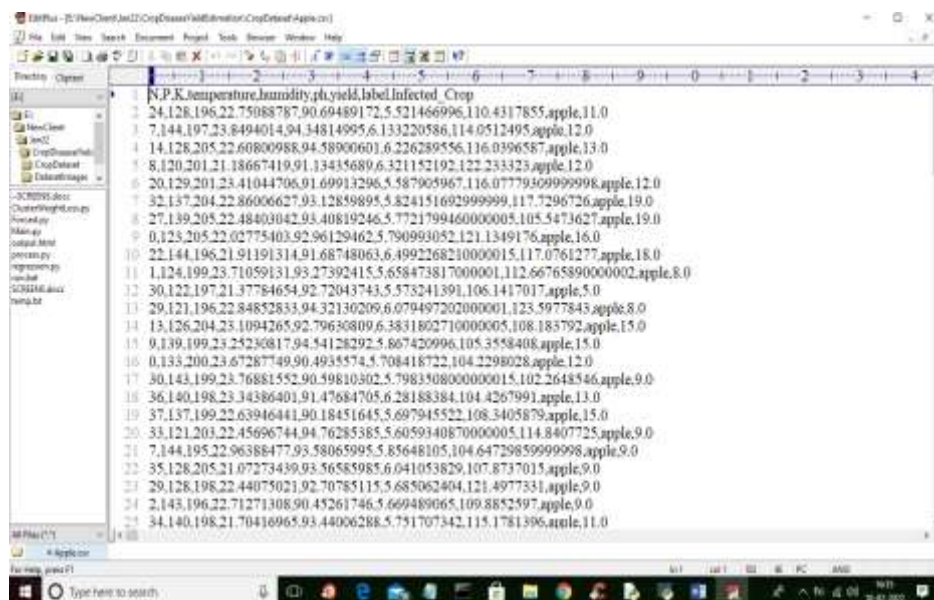
%% Preprocessing
img = imread('leaf_1.jpg');
[rows, cols, chns] = size(img);
img = img(:,:,1:3);

%% Clustering
[clusters, centers] = kmeans(img, 4, 'itermax', 100);

%% Weighting
weights = zeros(4, 1);
for i = 1:4
    weights(i) = sum(img == centers(i, :));
end;

%% Disease Detection
for i = 1:4
    [x, y, w, h] = find(img == centers(i, :));
    % Disease part extraction
end;
```

In above screen read red colour comments to detect and extract disease part from images and after applying above function on images will get final dataset with crop infected values as below screen In above screen we have 4 crop details and now open each file to view its content like below screen[8].



N	P	K	temperature	humidity	ph	yield	label	infected	crop
24.128	196.22	75088787.90	69489172.5	521466996.110	4317855	apple	11.0		
7.144	197.23	8494014.94	34814995.6	133220586.114	0512495	apple	12.0		
14.128	205.22	60800988.94	58900601.6	226289556.116	0396587	apple	13.0		
8.120	201.21	18667419.91	13435689.6	321152192.122	233323	apple	12.0		
20.129	201.23	41044706.91	69913296.5	587905967.116	077940999999	apple	12.0		
32.137	204.22	86006627.93	12859895.5	82415169299999	117.7296726	apple	19.0		
27.139	205.22	48403042.93	40819246.5	7721799460000005	105.5473627	apple	19.0		
0.123	205.22	02775403.92	96129462.5	79099052.121	1349176	apple	16.0		
22.144	196.21	91191314.91	68748063.6	4992268210000015	117.0761277	apple	18.0		
1.124	199.23	71059131.93	27392415.5	658473817000001	112.66765890000002	apple	8.0		
30.122	197.21	37784654.92	72043743.5	578241391.106	1417017	apple	5.0		
29.121	196.22	84852833.94	32130209.6	079497202000001	123.5977843	apple	8.0		
13.126	204.23	1094265.92	79630809.6	3831802710000005	108.183792	apple	15.0		
0.139	199.23	25230617.94	54128292.5	667420996.105	3558408	apple	15.0		
0.133	200.23	67287749.90	4935574.5	708418722.104	2298028	apple	12.0		
30.143	199.23	76881552.90	59810302.5	7983308000000015	102.2648546	apple	9.0		
36.140	198.23	34386401.91	47684705.6	28188384.104	4267991	apple	13.0		
37.137	199.22	63946441.90	18451645.5	697945522.108	3405879	apple	15.0		
33.121	201.22	45696744.94	76285385.5	6059340870000005	114.8407725	apple	9.0		
7.144	195.22	96388477.93	58065995.5	85648105.104	64729859999998	apple	9.0		
35.128	205.21	07273439.93	56585985.6	041053829.107	8737015	apple	9.0		
29.128	198.22	44075021.92	70785115.5	685062404.121	4977331	apple	9.0		
2.143	196.22	71271308.90	45261746.5	669489665.109	8852597	apple	9.0		
34.140	198.21	70416965.93	44006288.5	751707342.115	1781396	apple	11.0		

In above screen first row represents dataset column names and remaining are the dataset values and in above screen N refers to Nitrogen, K refers to potassium and P refers to phosphorus soil properties and remaining values are the normal environment data with infected and yield crop. By using above dataset to train all 4 machine learning algorithms and in below screen we can see all algorithm details

```

Python - E:\NewCloud Jan21\CropDiseaseYieldEstimation\train.py(372)
File Edit Format Run Options Window Help
[git,gitlab]([Python310\Python310 (64-bit)
git-2.42.0(64)
git-2.42.0(64)

#Function to run ENET algorithm
def runENET():
    global X, Y, X_train, X_test, y_train, y_test, ac, auc
    acc.clear()
    test_results = ['0', '0']
    best = float('-inf')
    for alpha in range(1, 1000):
        model = LinearRegression()
        model.fit(X_train, y_train)
        y_pred = model.predict(X_test)
        acc.append((alpha, model.score(X_test, y_test)))
        if model.score(X_test, y_test) > best:
            best = model.score(X_test, y_test)
            best_alpha = alpha

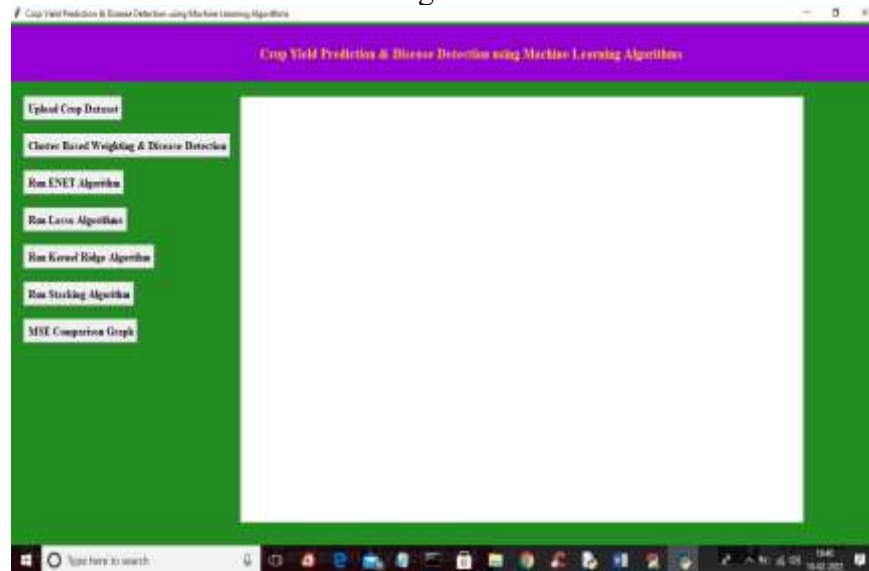
#Function to train LASSO algorithm
def runLASSO():
    global X, Y, X_train, X_test, y_train, y_test, ac, auc
    lasso = Lasso(alpha=1e-5, max_iter=1000)
    lasso.fit(X_train, y_train)
    y_pred = lasso.predict(X_test)
    acc.append((lasso.coef_, model.score(X_test, y_test)))

#Function to train Ridge algorithm
def runRidge():
    global X, Y, X_train, X_test, y_train, y_test, ac, auc
    ridge = Ridge(alpha=1e-5, max_iter=1000)
    ridge.fit(X_train, y_train)
    y_pred = ridge.predict(X_test)
    acc.append((ridge.coef_, model.score(X_test, y_test)))

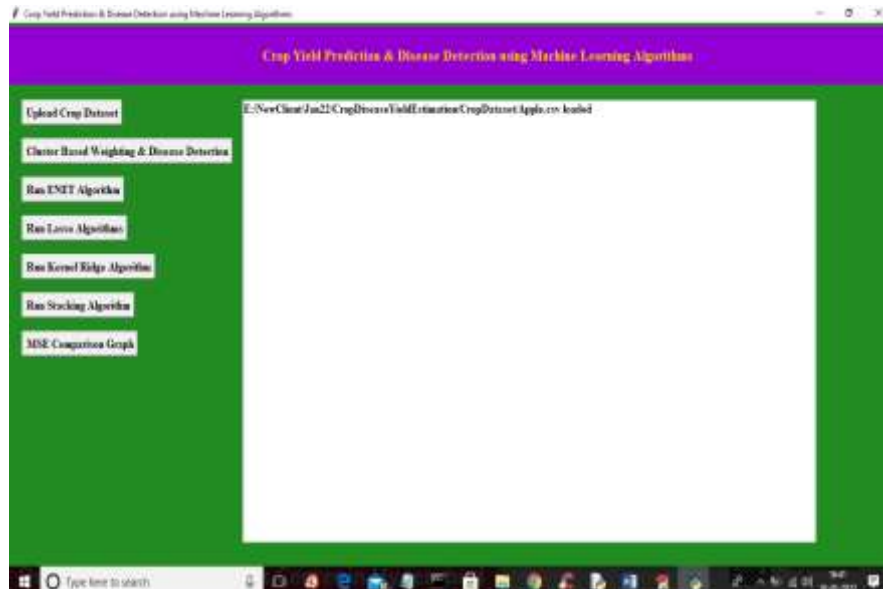
#Function to train stacking algorithm by using 3 different algorithms
def runStacking():
    global X, Y, X_train, X_test, y_train, y_test, ac, auc
    estimators = ['Ridge', 'Ridge', 'Ridge']
    stacker = StackingRegressor(estimators=estimators, final_estimator=LinearRegression())
    stacker.fit(X_train, y_train)
    y_pred = stacker.predict(X_test)
    acc.append((stacker.coef_, model.score(X_test, y_test)))

def main():
    height = 500
    data = ('ENET', 'Lasso', 'Ridge', 'Stacking')
    
```

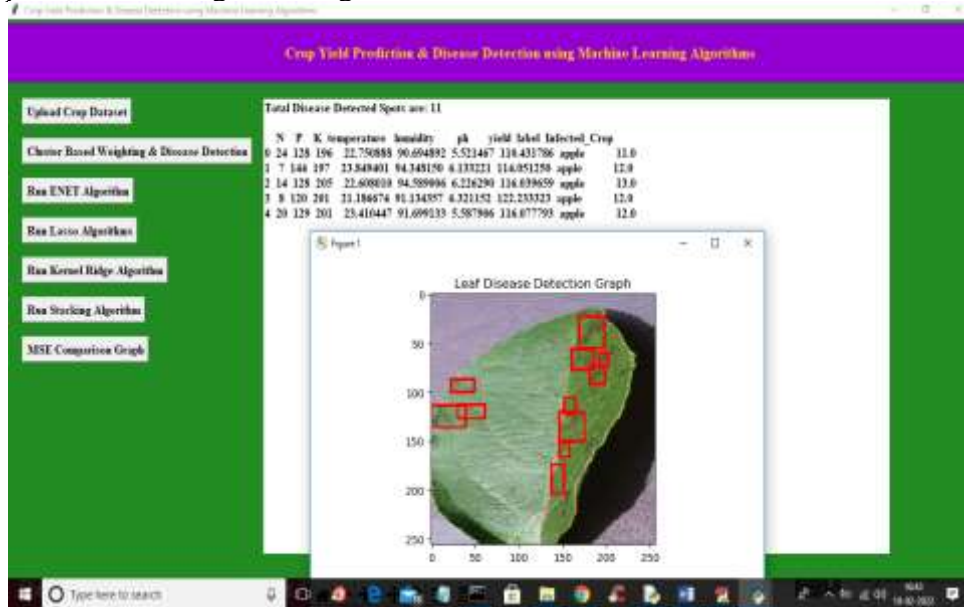
In above screen read red colour comments to know about all algorithms training. To run project double click on 'run.bat' file to get below screen



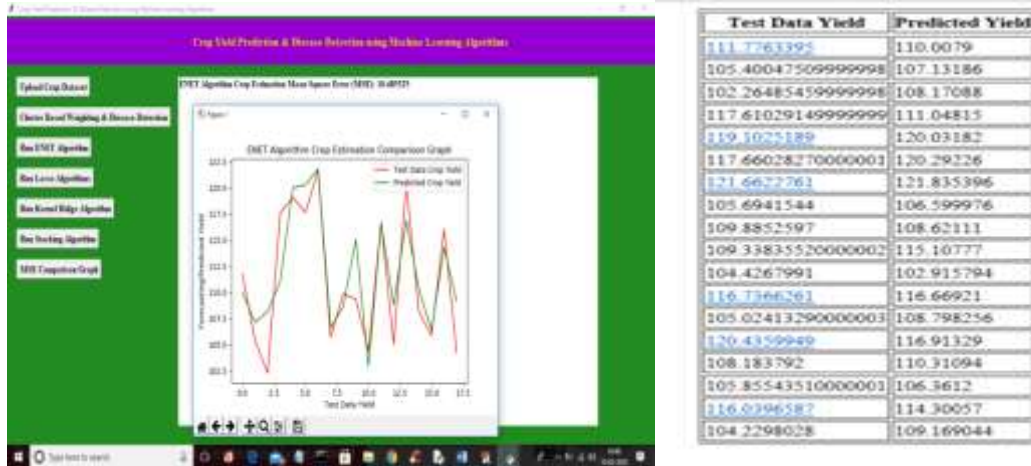
In above screen click on 'Upload Crop Dataset' button to upload dataset and to get below screen  
In above screen selecting and uploading 'Apple.csv' dataset file and then click on 'Open' button to get below screen



In above screen dataset loaded and now click on ‘Cluster Based Weighting & Disease Detection’ button to read images and then detect infected parts from the leaf and then add to dataset and for sample output I am showing one image of infected leaf like below screen



In above screen all infected leaf details added to dataset and then displaying total dataset details and then displaying one sample leaf showing disease part surrounded with red colour bounding boxes and now dataset is ready with crop and disease details and now close above image and then click on ‘Run ENET Algorithm’ button to get below output

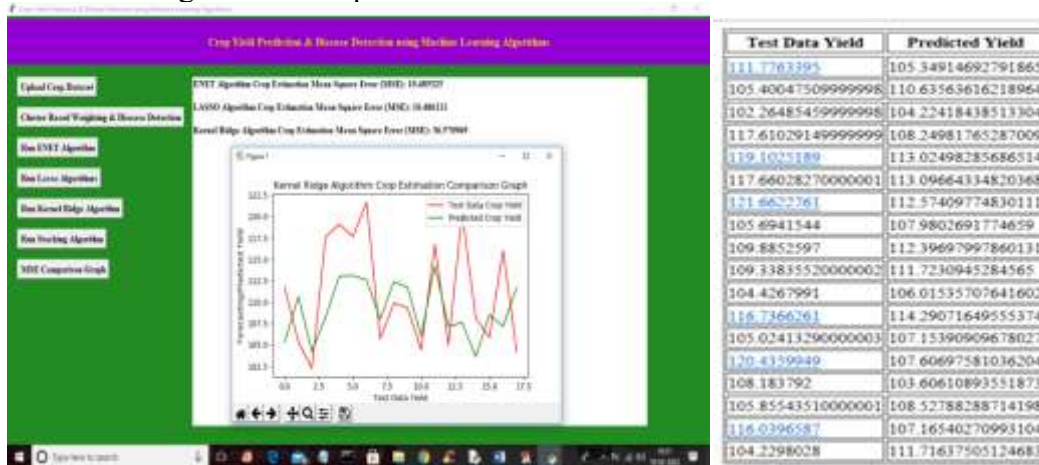


In above screen in text area we can see ENET got MSE value as 10 and in graph red line refers to TEST Crop DATA and green line refers to predicted crop yield and we can see in both lines very close difference is there so ENET is good in prediction and in browser we can see actual test data values and predicted values

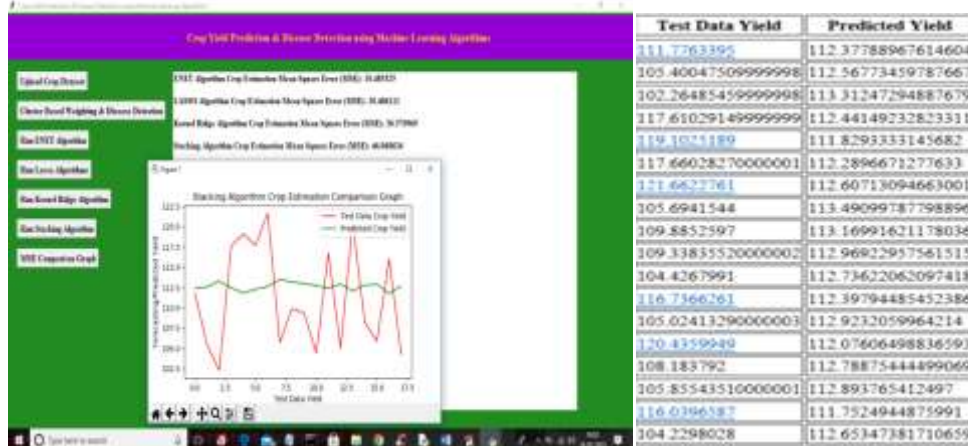
In above screen first column showing TEST data values and second column showing predicted crop yields in QUINTAL from ENET algorithm and now close above screen and graph and then click on ‘Run Lasso Algorithm’ button to get below output



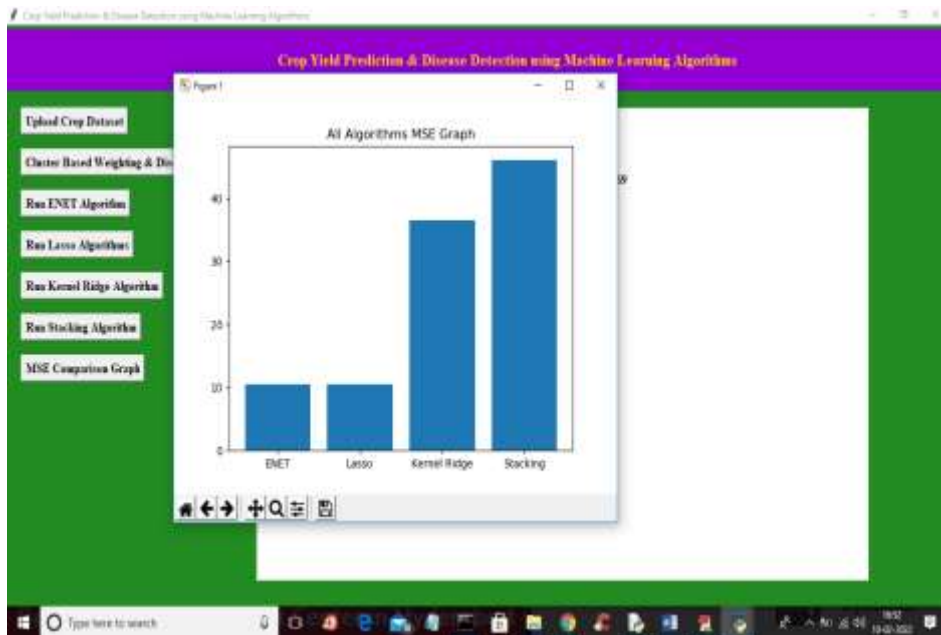
In above two screens you can see LASSO output screens and now click on ‘Run Kernel Ridge Algorithm’ button to get below output



In above two screens we can see Kernel Ridge output and now click on ‘Run Stacking Algorithm’ button to get below output



In above 2 screens we can see Stacking algorithm output and now click on ‘MSE Comparison Graph’ button to get below graph



In above graph x-axis represents algorithm names and y-axis represents MSE value and in all algorithms ENET and LASSO got less MSE values compare to other 2 algorithms and the lower the MSE the better is the algorithm. Similarly, you can upload other crop dataset and get output for those crops.

#### 4. CONCLUSION:

In the present work by combining plant leaf diseases image dataset with Crop Yield dataset to forecast or estimate future crop growth. To implement this project we have used various crop attributes such as Potassium, Phosphorous and Nitrogen properties of soil and environment data such as Humidity, temperature and PH values and then from crop leaf images we have extracted all and sum and average all infected leafs and included in CROP dataset[9-10].

Above process dataset will be trained with various machine learning algorithms such as Lasso, Kernel Ridge, ENET and Stacking of all 3 algorithms and then evaluate each algorithm performance in terms of MEAN SQUARE ERROR (MSE) and in all algorithms ENET and LASSO giving less MSE values.

We are using OPENCV inbuilt API'S to cluster leaf data based on colour values and then these values will be weighted to get spots whose part is not in green colour. Below code is used to extract or identify infected parts from the leaf [11-13].

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