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Mango Disease Detection by using Image Processing

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Abstract: Mango is the popular delicious fruit and cash crop. When diseases affect the mango plant there is significant decrease in the production of mango due to which farmers suffer in selling their yield. This problem motivated to develop the new techniques to detect and diagnose the diseases affecting the mango plant and devise the expert system to prevent those. To increase production and quality, it is very necessary to control all such harmful disease at the earlier stage. In our country most of the farmers are illiterate. So they cannot get correct information about disease. It requires an expert advisor or agricultural officer. But it is difficult for an expert advisor or an agricultural officer to reach at every farmer. In mentioned system For image enhancement "Imadjust" function is particularly used for contrast enhancement. This is a inbuilt function in MATLAB. Contrast enhanced is done by linearly scaling pixel values and features used are Contrast, Correlation, Energy, Entropy, Homogeneity, Cluster prominence, Cluster shade, Variance and Dissimilarity. Classifier used is SVM. The proposed system gives accuracy of 90% while testing on 92 samples.

Keywords: Mango disease, Production, Image Processing, diseases

I. INTRODUCTION

Now a days, production of mango fruit decreases because of climatic conditions and environmental concerns like heavy raining, high humidity, reduction in soil nutrients, diversity of associated diseases and disorder problems. Typically the detection of mango plant diseases is done by naked eye observation, which provides less accuracy. Low productivity of mango fruit is due to the various diseases affecting mango plant which are not recognized by the farmers as they are illiterate. In some rural areas farmers may have to go long distance to visit experts, this may be too expensive and time consuming and even farmers are unaware of diseases. Kisan call centers are available but do not service 24x7 hours and sometimes communication will be fail. Sometimes response getting from the agricultural officer is delayed. Hence proper disease detection is not done within short time which results in reduction of mango fruit production. This paper gives basic idea about detection and recognition of different diseases occurred on various parts of the mango plant like stem, fruit, leaves, branch etc. also proper preventive care is provided for the affected region of the mango plant so that production of mango fruit will increase as well as quality of the mango fruit will improve.

II. LITERATURE REVIEW

The main reason why we have chosen this idea of detecting mango diseases is to support production of mango fruit in India, that is depleted due to various diseases. This paper illustrates how different diseases affect mango fruit productivity. The main diseases that affect mango fruit productivity are Anthracnose, Powdery Mildew and Red Rust according to paper thus we have chosen this four diseases along with a subsidiary disease called Black Banded. This paper also illustrates preventative ways and cure for diseases [1]. This paper proposes MATLAB based disease detection system. This system proves it's ability to detect leaf based diseases though there are many major diseases that affect the productivity of mango crop. For example Black Band which majorly occurs on barks or stem and still affect productivity. Also some diseases are also there which occurs on fruit itself. Accuracy is also main constrain that needs improvement.

First step of digital image processing is Image Preprocessing which is one of the main steps for retaining originality in an image that may be degraded due to several noises or due to other causes. This paper is a review of several Image Preprocessing techniques that can be employed [2].

Image segmentation is important for choosing proper area of interest in an image. This is important to reduce processing time that may be required to process rather entire image. This paper shows new segmentation technique called Fast and Robust Fuzzy C-Means algorithm. The beauty of this algorithm is that it automatically chooses cluster that contains disease. Also it contradicts showing advantages over traditional FCM algorithm [3].



This paper presents review on features to be extracted. Features that can be extracted from image may of various types like global or local. In that there are subcategories like texture based or color based features etc. [4] [5].

The best and easy way to calculate texture based features in MATLAB based system is by using GLCM(Gray level Co-occurrence Matrix).MATLAB has in-built function for drawing GLCM. This paper provides equations for calculating various texture based features from GLCM matrix. GLCM is matrix function of angle and distance thus computational cost of GLCM is very high. This paper presents directional analysis of gray level images using GLCM. According to this paper, choosing 0 direction yields better results [6].

In feature extraction, for successful disease detection it is necessary to make proper selection of features that best describes the given disease is important. This paper illustrates different features selection algorithms and feature scoring methods. According to this paper Sequential Forward Selection is better method for feature subset selection and reducing dimensionality [7].

III.PROJECT OVERVIEW

In this system, farmer capture picture of the infected leaf, fruit, stem, flower and send it to system via internet. Once system receive the image of infected part of mango plant that image will be input to the system .After this, technical processes will be applied on the image such as Image pre-processing which includes steps like resizing of images, image cropping, image enhancement etc. After that image Segmentation is done which divide the image into small parts for better results. For image segmentation Fast and Robust Fuzzy-C means clustering algorithm is used .Feature extraction is to extract the features from the affected part of the mango plant. Feature Extraction is done by Gray Level Co-occurrence Matrix (GLCM) algorithm. SVM classifier is used to classify the type of disease according to the features extracted. After applying all this technical processes, at the output, system displays name of the disease and preventive care according to the disease identified. This information is send to the farmer as a feedback. In the proposed methodology, several steps are carried out like image acquisition, image pre-processing, image segmentation, feature extraction and classification of images. In the present work several frames, i.e., Images are captured by the farmer with the help of 8 MP camera. Each image is pre-processed in order to enhance and remove the noise from it. The segmentation techniques are applied to detect the affected area. Further, the Feature and Classification are done to extract the defected part of the image and to calculate affected area of the leaf. In proposed system MATLAB R2013a is used and GUI is made on same platform for better user experience.

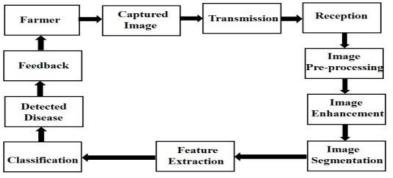
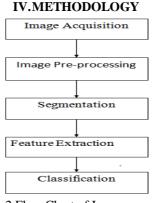
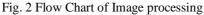


Fig. 1 Block diagram of mango disease detection







A. Image Acquisition

Image acquisition is first step in digital image processing. This captures the image through 8 Mega Pixel camera and stores it in digital media for further MATLAB operations. In our work, using digital camera we captured healthy and diseased images of leaves, fruits and stems.

B. Image Pre-Processing

When image resizing is done then interpolation occurs. Image resizing is necessary when there is need to increase or decrease the total number of pixels in image. If number of pixels are less then processing speed will be faster and result is obtained within short period.

Image enhancement is done to enhance the image to get correct region of interest. The main aim of image enhancement is to improve visual quality and brighten the information contents present in the original image before actual processing.in our proposed work, image enhancement is done to easily detect and recognize disease affected area so that proper result will obtain.



Fig 3 Image enhancement

C. Image Segmentation

For image segmentation fast and robust fuzzy c-means clustering algorithm is used. As fuzzy c-means clustering (FCM) algorithm is sensitive to noise, local spatial information is often introduced to an objective function to improve the robustness of the FCM algorithm for image segmentation. The introduction of local spatial information often leads to a high computational complexity, arising out of an iterative calculation of the distance between pixels within local spatial neighbors and clustering centers. An improved FCM algorithm based on morphological reconstruction and membership filtering (FRFCM) that is significantly faster and more robust than FCM, is proposed in this project to address this issue. FRFCM algorithm is simpler and significantly faster.

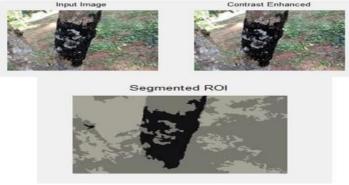


Fig. 4 Image segmentation

D. Feature Extraction

Feature extraction is important step in image processing that includes extraction of features from given image to get meaning of given image. Goal of the features extraction is to extract correct, unique features that distinguish a image from others in given set. There are many types of features including shape, color based or texture based etc. (There might a lot numbers of features other than this). For detection of diseases, here textural features are considered. This is because we found discrepancy in textures of chosen diseases. For measuring various features Statistical method is employed. A popular method is to use GLCM (Gray-Level Co-Occurrence Matrix). The GLCM functions characterize the texture of an image by calculating how often pairs of pixel. Following are the extracted features from GLCM matrix:



1) Contrast: Contrast can be simply explained as difference between maximum and minimum intensity of pixel. It can be formulated as,

$$\text{Contrast} = \sum_{i,j=0}^{N-1} P(i,j) \times (i-j)^2$$
(1)

2) Correlation: A Correlation shows how pixel is correlated to its neibourhood. Its value ranges between -1 to 1.

Correlation=
$$\sum_{i,j=0}^{N-1} P(i,j) \times \frac{(i-\mu) \times (j-\mu)}{\sigma^2}$$
(2)

3) Energy: Energy measures uniformity with squared elements in GLCM. Range is in between 0 to 1 value of energy is 1 for constant image.

Energy=
$$\sum_{i,j=0}^{N-1} P(i,j)^2$$
 (3)

4) Entropy: Entropy measures disorder in an image and it achieves largest value when all p matrixes are equal.

Entropy =
$$\sum_{i}^{N-1} \sum_{j}^{N-1} P(i, j) \times \log[P(i, j)]$$
(4)

5) Homogeneity: Homogeneity gives value that measures closeness of distribution of elements in GLCM to GLCM diagonal.

Homogeneity=
$$\sum_{i,j=0}^{N-1} P(i,j) / R$$
(5)

6) *Cluster prominence:* Cluster prominence measures asymmetry. When value of this feature is high, image is less symmetrical and vice versa.

Cluster Prominence=
$$\sum_{i=0}^{N-1} \sum_{j=0}^{N-1} P(i,j) \times (i-j-\mu)^4$$
 (6)

7) Cluster shade: Cluster shade similar to cluster prominence measures skewness.

Cluster shade=
$$\sum_{i=0}^{N-1} \sum_{j=0}^{N-1} (i - j - \mu)^3 \times P(i, j)$$
 (7)

8) Variance: It measures dispersion of gray level distribution.

Variance=
$$\sum_{i=0}^{N-1} \sum_{j=0}^{N-1} (i - \mu)^2 \times P(i, j)$$
 (8)



9) Dissimilarity: It measures dissimilarity in successive pixels in image.

Dissimilarity=
$$\sum_{i,j=0}^{N-1} P(i,j) \times |i-j|$$
(9)

E. Classification

Classification is most important part in image processing technique which is based on the classifiers. The main function of the classification is to correctly identify the value of a variable that are of discrete class. Classification is done in plant disease detection which depends on whether the image is infected by disease or not. In our proposed work SVM classifier is used for classification. Support Vector Machine is linear model for classification problem and regression problems. SVM algorithm can creates line or a hyperplane which separates the data into classes. SVM's do a separating line (or hyperplane) between data of two classes. SVM is a algorithm that takes the data as input and outputs a line that separates those classes if possible. Suppose the dataset as shown below and you need to classify the blue triangle from the yellow ellipses. So the task is to find an ideal line that separate this dataset in two classes (blue and yellow). We have infinite lines that can separate these two classes as shown in fig. According to SVM algorithm find the points closest to the line from both the classes. These points are called as support vectors. Then compute the distance between the line and the support vectors. This distance is called as margin. The hyperplane for the margin is maximum in the optimal hyperplane. A hyperplane is an n-dimensional Euclidean space is a flat, n-1 dimensional subset of that space that divide the space into two disconnected parts. SVM tries to make a decision boundary in such a way that separation between the two classes is as wide as possible.

V. RESULTS AND DISCUSSION

The proposed system has built to detect 4 diseases named Anthracnose, Powdery Mildew, Black Banded, Red Rust and to provide solution over them .The GUI and programming is made using MATLAB. The various features selected has following values.

Fig. 6 shows Anthracnose disease on mango leaf. By calculating parameters using GLCM algorithm we made training database of 50 samples and determine average value of parameters i.e. Autocorrelation=17.37, Entropy=0.862, Energy=0.49, Variance=17.48, Contrast=0.506 etc. By comparing image sent by farmer with the trained database we get detection of Anthracnose disease on mango plant. Also solution for Anthracnose disease is displayed on the screen.

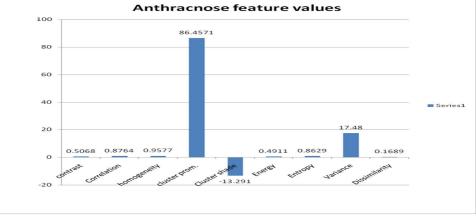


Fig. 5 Features values for Anthracnose



hpd Image	Contrast Enhanced		Extracte	d Features		Help Dialog - D	GUI	3 ×
		Autocorrelation		Vanance	17 4817	Antheorem	Authra	cnose
		Contrast	0.506801	Sum Average	7.89283		Carbendazi	1gm/ltr
		Correlation	0.876441	Sum Variance	57 6776	Result		
Select Image	Enhance Contrast	Cluster Prominence	86.4571	Sum Entropy	0 823071	Anthracnose	Thiophanate	2gm/ltr
Segmen	ted ROI	Cluster Shade	-13.2901	Diff Variance	0 506801		Propineb	2gm/ltr
	Card a	Dissimilarity	0.168934	Diff Entropy	0.218695			
	2 - Zur	Energy	0.491147	Max Probability	0.62065	Solution		
and a	A1 2	Entropy	0.86207	Homogeneity	0.957767			
Segme	nt Image					BACK		

Fig. 6 Anthracnose disease detection on mango leaf

Input Image Contrast Enhanced					Metp Dialog	GUI3 - 🗆 🗙
			d Features		Arthucrose	Anthracnose
	Autocorrelation	0.286351	Vatiance Sum Average	16.2204 7.63258		Carbendazi 1gm/ltr
Select Image Enhance Contrast	Correlation Cluster Prominence	0.91968 98.3784	Sum Vanance Sum Entropy	54.8964 0.700439	Authracoose	Thiophanate 2gm/ltr
Segnemed ROJ	Cluster Shade	18.3252	Diff Variance	0.288351		Propineb 2gm/ltr
	Dissimilarity Energy	0.0954564	Diff Entropy Max Probability	0.141002	Solution	
Xa mark	Entropy	0 722488	Homogeneity	0.976137		
Segmented Image					BACK	

Fig. 7 Anthracnose disease detection on mango fruit

Input Image	Contrast Enhanced					Thelp Dialog - Dialog	GUI3	
Atom	A	Autocorrelation		d Features Variance	14 5403	OK	Anthraci	iose
	de.		0.0017892	Sum Average	7,41484	Result	Carbendazi	1gm/ltr
Select Image	Enhance Contrast	Correlation Oluster Prominence	0 949900	Sum Variance Sum Entropy	48.0895	Anthracnose	Thiophanate	2gm/ltr
Segnette	180		4.12052	Diff Vanance	0.0917892		Propineb	2gm/ltr
	-	Energy	0.520409	Max Probability		Solution		
2	and the second s	Entropy	0.757926	Homogeneity	0.984702			
Segmente	ed Image					BACK		

Fig. 8 Anthracnose disease detection on mango flower

Fig. 10 shows Powdery Mildew disease on mango leaf. By calculating parameters using GLCM algorithm we made training database of 50 samples and determine average value of parameters i.e. Autocorrelation=16.094, Entropy=1.283, Energy=0.308, Variance=16.13, Contrast=0.203 etc. By comparing image sent by farmer with the trained database we get detection of Powdery Mildew disease on mango plant. Also solution Powdery Mildew disease is displayed on the screen.



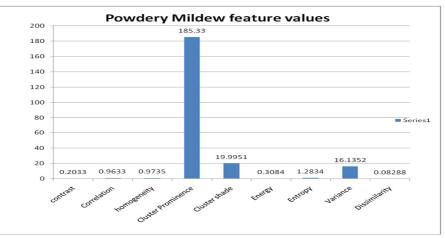
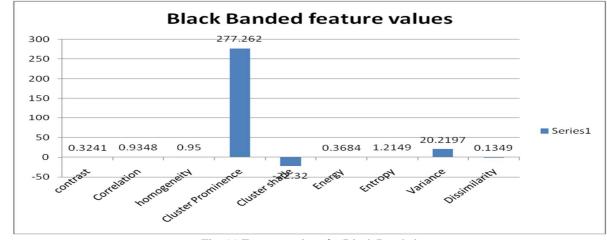


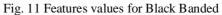
Fig 9 Features values for Powdery Mildew

Input Image	Input Image Centrast Enhanced		Extracted	d Features	Í. h	🖸 Help Dialog 🕒 🗆 🗙	0	GUI3 - 🗆 🗙
-	- Carl	Autocorrelation	16.0947	Variance	16.1352	Powdery Mildew		Powdery
1111	11/18	Contrast	0.203385	Sum Average	7.33366	OK		0.1.1
		Correlation	0.963063	Sum Variance	47.7584	Result		Sulphur 0.2%
Select Image	Enhance Contrast	Cluster Prominence	185.303	Sum Entropy	1.25428	Powdery Mildew		Hexaconazo 0.05%
Segment	ed ROI	Cluster Shade	19.9951	Diff Variance	0.203385			Triademefon 0.1%
	·	Dissimilanty	0.0826891	Diff Entropy	0.20668			21300000000
7.26	find the	Energy	0.308469	Max Probability	0.358058	Solution		Carbendazin 0.1%
A		Entropy	1.28346	Homogeneity	0.973599			
Segmen	nt Image					BACK		

Fig.10 Powdery mildew disease detection on leaf

Fig. 12 shows Black Banded disease on mango stem. By calculating parameters using GLCM algorithm we made training database of 50 samples and determine average value of parameters i.e. Autocorrelation=20.17, Entropy=1.214, Energy=0.368, Variance= 20.21, Contrast=0.324 etc. By comparing image sent by farmer with the trained database we get detection of Black Banded disease on mango plant. Also solution for Black Banded disease is displayed on the screen.







Input Image	Contrast Enhanced		_				GUB	- C X
			Extracted	d Features		Help Dialog - =		_
8		Autocorrelation	20 1722	Variance	20.2197	Back Banded	Black	
		Contrast	0.324142	Sum Average	8.44874	OK	Endosulfan	15ml
-		Correlation	0 934886	Sum Variance	62.5483	Result		
Select Image	Enhance Contrast	Cluster Prominence	277.262	Sum Entropy	1.17436	Black Banded	Carbany	20ml
Segment	ed ROI	Cluster Shade	-22.3276	Diff Variance	0.324142		Imidachlopr	3ml
		Dissimilarity	0.134589	Diff Entropy	0.26627			
1 A.S.		Energy	0.368492	Max Probability	0.510325	Solution	Di methoate	10ml
	2 mg	Entropy	1.21492	Homogenesty	0.959481		Cypermethri	3ml
Segmen	nt Image					BACK		

Fig. 12 Black Banded disease detection on stem

Fig. 14 shows Red Rust disease on mango plant. By calculating parameters using GLCM algorithm we made training database of 50 samples and determine average value of parameters i.e. Autocorrelation=25.03, Entropy=1.264, Energy=0.314, Variance= 25.10, Contrast=0.274 etc. By comparing image sent by farmer with the trained database we get detection of Red Rust disease on mango plant. Also solution for Red Rust disease is displayed on the screen.

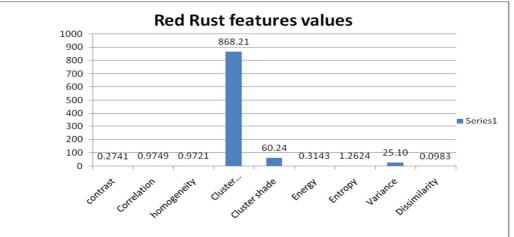


Fig.13 Features values for Red Rust

Input Image	Contrast Enhanced		Extracted	d Features	1	🛛 Help Dialog 🔶 🗆 🗙	GUB - 🗆 🗙
		Autocorrelation	25.031	Variance	25.1071	Red Rust	Red Rust
CHAIR SEX	ALC: LA	Contrast	0.274203	Sum Average	8.87662	OK.	Bordeaux 1%
		Correlation	0.974955	Sum Vanance	79.9937	Result	
Select Image	Enhance Contrast	Cluster Prominence	868.689	Sum Entropy	1.23529	Red Rust	Copper 2.5ltr
Segmente	ed ROI	Cluster Shade	60.2442	Diff Variance	0.274203		
	2.44	Dissimilarity	0.0983762	Diff Entropy	0.190982		
	1.3	Energy	0.314351	Max Probability	0.382123	Solution	
1		Entropy	1.26248	Homogeneity	0.972119		
Segmen	t Image						
oeginer	it mage					BACK	

Fig.14 Red Rust disease detection on leaf



Above fig 16.shows healthy mango. By calculating parameters using GLCM algorithm we made training database of 50 samples and determine average value of parameters i.e. Autocorrelation=39.92, Entropy=1.047, Energy=0.401, Variance=27.29, Contrast=0.1343 etc. By comparing image sent by farmer with the trained database we get detection of healthy mango. We take total 10 samples of healthy parts of mango and out of them 7 samples are perfectly detected for healthy parts of mango.

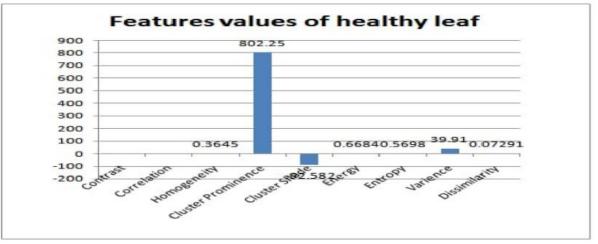
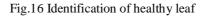


Fig.15 Features value of healthy leaf

Input Image	Contrast Enhanced		Extracted	d Features		Help Dialog X
	ALC: NO	Autocorrelation	24.0686	Variance	24.0454	OK .
ALC NOT		Contrast	0.255132	Sum Average	8.768	
		Correlation	0.974369	Sum Variance	76.9004	Result
Select Image	Enhance Contrast	Cluster Prominence	461.461	Sum Entropy	1.20172	Healthy
Segmen	ted ROI	Cluster Shade	22.7937	Diff Variance	0.255132	Stanta a
the second s		Dissimilarity	0.0686734	Diff Entropy	0.149274	
		Energy	0.327218	Max Probability	0.407552	Solution
and a		Entropy	1.22003	Homogeneity	0.982798	
Segmen	ited Image					BACK



Input Image	Contrast Enhanced		Extracted		Help Dialog	
		Autocorrelation	39.9239	Variance	39.9117	Healty OK
-	-	Contrast	0.364583	Sum Average	12.0236	
		Correlation	0.954019	Sum Variance	146.913	Result
Select Image	Enhance Contrast	Cluster Prominence	802.258	Sum Entropy	0.559757	Healthy
Segne	ted ROI	Cluster Shade	-92.5885	Diff Variance	0.364583	
		Dissimilarity	0 0729167	Diff Entropy	0 076133	
1	4	Energy	0.668471	Max Probability	0.795067	Solution
	A 2	Entropy	0.569863	Homogeneity	0.987847	
Segme	ent Image					BACK

Fig.17 Identification of healthy fruit



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VI.CONCLUSIONS

The proposed system enables us to detect and recognize diseases affected on mango plant and to provide appropriate preventive care and solution for it. The system makes the detection of diseases easier by complete automation and notification of the affected diseases and remedies are sent to the farmer promptly. This system will increase productivity and improve quality of mango fruit. In given system processing speed of segmentation is 3 seconds and results will be displayed after classification in 0.1 seconds. The overall system will worked in 5 seconds.

VII. ACKNOWLEDGMENT

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