

## AN INVESTIGATION ON CONVOLUTIONAL NEURAL NETWORKS FOR THE DETECTION AND IDENTIFICATION OF PLANT LEAF DISEASES

<sup>1</sup>Preethi Dashyam, <sup>2</sup>Anitha Thotha, <sup>3</sup>Ravi Shankar Sunkaraboina, <sup>4</sup>Vupeti Rekha

<sup>1,2,3</sup>Assistant Professor, <sup>4</sup>UG Student, <sup>1,2,3,4</sup>Department of Computer Science Engineering, Rishi MS Institute of Engineering and Technology for Women, Kukatpally, Hyderabad.

### ABSTRACT

This job primarily used some Deep Learning techniques to identify the early stages of a plant's disease. We are aware that India's agricultural sector generates a substantial amount of revenue. Bug damage to plants and harvests has an impact on the nation's horticulture production. Sickneses and irritant infestations practically cost the world's crop produce 40% of its output. In farming, the early location of illnesses is important for a good crop yield. The Machine Learning methodology is one of the methods for detecting plant diseases. Our approach progresses toward the notion of deep learning-based brain networks.

### INTRODUCTION

India is a nation where a larger portion of the population is heavily dependent on the countryside. The plant's vulnerability to diseases is one of the main problems with gardening. By continuously inspecting plants over a long period of time, experts can identify objective information regarding a plant's diseases. Illnesses do have negative impacts on plants, which have an impact on their normal growth. In terms of the farming system, India has come in eighth. According to the evaluation, 56.6% of people work mostly in agriculture.

Horticultural innovation has spread quickly across life in multiple directions. Ranchers started using extended social practices and agricultural contributions with work-escalated initiatives after the green upset to boost the harvest yield potential per unit of land.

Plants act as a spine to support the climate. Plants truly do experience the ill effects of sicknesses, which influence the typical development of plants. These illnesses influence total plant development including leaf, bloom, stem, and organic product. Discovery of such plant sicknesses is a significant assignment to perform. More often than not the it are tedious and bulky to exist ways to deal with sickness distinguishing proof. Thus, to screen plant sickness at a beginning phase, the utilization of a few programmed techniques can be very gainful.

Tomato is the most widely recognized vegetable utilized across India. The tomato crop development region in India ranges around 3,50,000 hectares roughly and the creation amounts generally summarize to 53,00,000 tones, making India the third-biggest tomato maker on the planet. Sickness impacted plants comprise 10-30% of absolute harvest misfortune.

This work supplements agriculturists to merge the instinctive choice with respect to the acknowledgment of plant infection by utilizing profound learning conventional model carried out through sensor flow. The calculation utilized at the center of the profound learning model is the Convolutional Neural Networks (CNN or ConvNet) is a kind of feed-forward counterfeit brain network in which the availability design between its neurons is roused by the association of the creature visual cortex. The model characterizes pictures of any plant into sick and solid ones. The dataset utilized for this work is comprised of tomato plants photos, subsequent to expanding those utilizing pivots at different points.

## LITERATURE SURVEY

Sushil R. Kamlapurkar's [1] proposed approach is used to identify plant leaf diseases. In this paper, the Gabor filter is applied to feature extraction, and after performing feature extraction, classified the plant leaf diseases using an artificial neural network. The main disadvantage of this paper is computational complexity.

Surender Kumar and Rupinder Kaur [2] proposed another approach to identifying plant diseases. The infected part is segmented based on a threshold value, based on the edge, based on the similarities and dissimilarities. After performing segmentation, the feature is extracted using the color occurrence method, grey level co-occurrence matrices, and classified the crop diseases using the linear and non-linear filters.

Usama Mokhtar et al[3] have proposed an efficient method that identifies whether a tomato leaf is healthy or infected. The image given as input was first preprocessed by removing the background and the noise present was eliminated with the help of the erosion technique. Gray level Co-occurrence Matrix (GLCM) was used for texture feature extraction from the enhanced image. Support Vector Machine (SVM) classifier was trained using different kernel functions and the performance has been evaluated using the N-fold cross-validation technique. The proposed system had achieved an accuracy of 99.83% using the linear kernel function with the SVM classifier. Even though the obtained accuracy is high, it is not sufficient enough to predict or differentiate between healthy or diseased leaves. Also, the type of disease was not identified.

In [4], a novel cucumber leaf disease detection system was presented based on convolutional neural networks. Under the fourfold cross-validation strategy, the proposed CNN-based system achieved an average accuracy of 94.9% in classifying cucumbers into two typical disease classes and a healthy class. The experimental results indicate that a CNN-based model can automatically extract the requisite classification features and obtain performance.

## DIFFERENT DISEASES ON TOMATO LEAF

### 1. Tomato Mosaic

Tomato Mosaic Virus (ToMV) is a plant pathogenic infection. It is seen as overall and influences tomatoes and numerous different plants. The foliage of impacted tomato plants shows mottling, with exchanging yellowish and hazier green regions, the last option frequently seeming thicker and raised giving a rankle like appearance



### 2. Early Blight

Early Blight is a typical tomato infection is brought about by the parasite *Alternaria solani*. It can influence practically all pieces of the tomato plants, including leaves, stems, and organic products. The plants may not bite the dust, yet they will be debilitated and will set less tomatoes than ordinary. Early Blight by and large goes after more seasoned plants, however it can likewise happen in seedlings.



### 3. Bacterial Wilt

Bacterial wilt is an overwhelming nursery sickness, making tomatoes and other nightshade vegetable plants shrivel and pass on out of nowhere and with very little advance notice. It is almost difficult to treat, however there are steps you can take to forestall its spread. Bacterial wilt is a dirt borne sickness brought about by the microscopic organisms *Ralstonia solanacearum*.



## METHODOLOGY

The initial phase in fostering the model is to gather more information to work with, investigate the information completely and get the different boundaries/credits in it. In this stage, the pictures of tomato leaves are gathered from the web or the ranch. The pictures gathered may contain sound leaves and sick leaves. The pictures of the leaves are gathered with various directions. The dataset is ready by isolating it into different classifications. The pictures are characterized into great, terrible, and most exceedingly awful in view of their highlights. The pictures are saved in .jpg and .png designs.

The model is made utilizing 64 info neurons and comprises of 2 halfway layers. The 64 info neurons give 64 neurons to the main layer. Further, the 32 neurons are utilized to create the necessary result. The 64 sources of info are preoccupied from the elements of the prepared pictures of the dataset. The brain network model is created utilizing the Keras and TensorFlow capacities.

Finally, assessing the model depends on the prepared model is tried utilizing test pictures. The library capacities utilized in python are Cv2, Os, Matplotlib, Glob, Numpy. The libraries used to

to foster a Neural organization are Keras, TensorFlow. The capacities to actuate the Neural Network are ReLU.

The model which is utilized for the picture recognizable proof is the Histogram of Oriented Gradients. The histogram of situated inclinations is a component descriptor utilized in PC vision and picture handling for object location. The strategy includes events of slope direction in confined segments of picture. The means engaged with the execution of HOG are preprocessing, Gradient calculation, Orientation binning, Block standardization, Object acknowledgment.

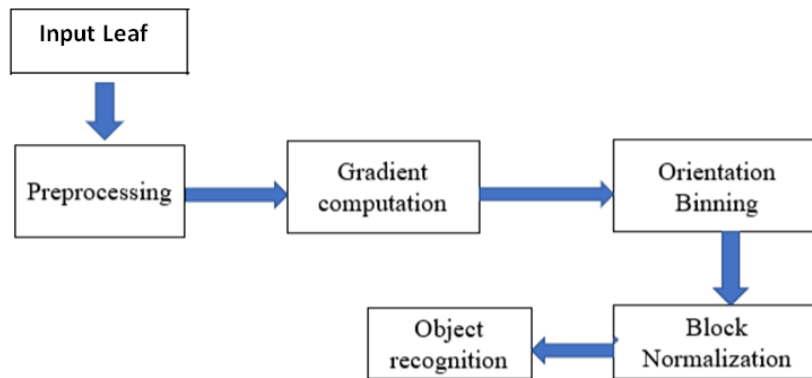
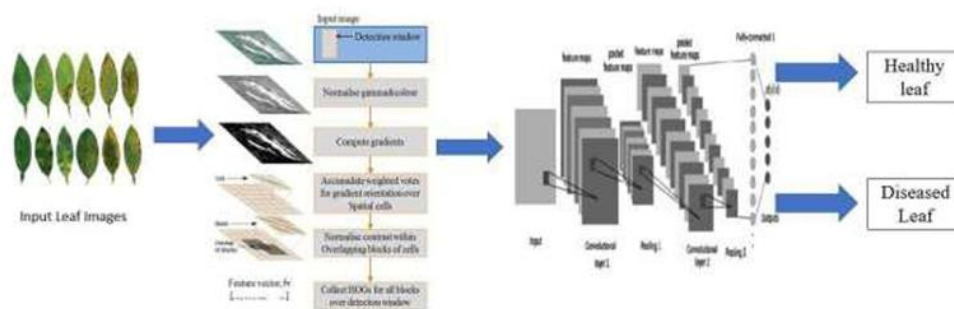


Fig: HOG function implementation

### MODELING AND ANALYSIS



### RESULTS

Input:

Here we take the dataset of some tomato leaves and the dataset is categorized into 3 parts such as good, bad, worst. Each categorized part consists of some tomato leaves. The good part consists of healthy leaves, the bad part consists of lightly affected leaves and the worst part consists of fully affected leaves.

Output:

In the beginning, it will provide how many healthy and unhealthy leaves are there. Our neural network takes the input as a leaf image and produces the percentages of leaf features such as bacterial spots, healthy leaf, how much it gets affected by the disease.



Fig: Illustrates the output of a Healthy Leaf leaf.

Fig: Illustrates the output of diseased and infected leaf.

## CONCLUSION AND FUTURE SCOPE

Infestations that cause discomfort and illnesses practically cost the global agriculture industry 40% of its productivity. In farming, a healthy crop production depends on the early detection of infections. One technique for identifying plant diseases is the machine learning approach. Our method moves closer to the idea of brain networks based on deep learning. Different learning rates and optimizers could also be used for experimenting with the proposed model as a part of future work. It could also include experimentation with newer architectures for improving the performance of the model on the train set.

## REFERENCES:

1. Sushil. R. Kamapurkar, "Detection of plant leaf diseases using image processing approach", International journal of scientific and research publication, volume 6, issue2, February 2016.
2. Surrender Kumar Rupinder Paur, "Plant disease detection using image processing-a review", International Journal of computer application, volume 124-no 16, August 2015.
3. Usama Mokhtar et al. "Tomato leaves diseases detection approach based on support vector machines", Computer Engineering Conference (ICENCO), 2015 11th International. IEEE. 2015, pp. 246–250.
4. S. D. Khirade and A. B. Patil. "Plant Disease Detection Using Image Processing", 2015: International Conference on Computing Communication Control and Automation, February 2015, pp. 768–771, DOI:10.1109/ICCUBEA.2015.153.
5. Kawasaki, Y. Uga, H. Kagiwada, S. Iyatomi, "Basic study of automated diagnosis of viral plant diseases using convolutional neural networks", The 12th International Symposium on Visual Computing, Las Vegas, NV, USA, pp. 638–645, 12–14 December 2015.
6. Sladojevic. S, Arsenic. M, Andela. A Culibrk. D, Stefanovic, "Deep neural networks-based recognition of plant diseases by leaf image classification", Computer Intell. Neurosci. 2016, pp. 1- 11.
7. Malvika Ranjan, Manasi Rajiv Weginwar, Neha Joshi, A.B.Ingole, "Detection and classification of leaf diseases using artificial neural network", International journal of technology and applications, May-June 2015.
8. Loyce Selwyn Pinto, Argha Ray, M.Udhayeswar Reddy, Pavithra Perumal, Aishwarya, "Crop diseases classification using texture analysis", IEEE international conference on recent in information communication technology, May 20-21, 2016, India.
9. Jagadeesh. D, Pujari Rajesh, Yakkundimath, Abdulmunaf S.Byadgi, "Identification and Classification of Fungal disease Affected on Agriculture/Horticulture Crops using Image Processing Techniques", 2014 IEEE International Conference on Computational Intelligence and Computing Research.
10. Jayamala K. Patil, Raj Kumar, "Advanced in image processing for detection of plant diseases", Journal of Advanced Bioinformatics Applications and Research ISSN 0976-2604 Vol 2, Issue 2, June-2011, pp 135-141.
11. H Sabrol and K Satish. "Tomato plant disease classification in digital images using classification tree", Communication and Signal Processing (ICCSP), 2016 International Conference on. IEEE. 2016, pp. 1242–1246.

12. Usama Mokhtar et al. "Tomato leaves diseases detection approach based on support vector machines", Computer Engineering Conference (ICENCO), 2015 11th International. IEEE. 2015, pp. 246–250.
13. Monteiro ST, Kosugi Y, Uto K, Watanabe E, "Applying Hyperspectral imagery as an intraoperative visual aid tool", Proceeding of the Fourth IASTED International Conference on Visualization, Imaging and Image Processing, Marbella, Spain, pp. 483-488.
14. Paola JD, Schowengerdt RA, "A Review and analysis of backpropagation Neural Networks for classification of remotely-sensed multispectral imagery", International Journal of Remoting Sensing, 1995, Vol. 16, Issue 16, pp. 3033-3058.
15. Sladojevic. S, Arsenic. M, Andela. A Culibrk. D, Stefanovic. D, "Deep Neural Networks based recognition of plant diseases by leaf image classification", Comput. Intell. Neurosci. 2016, pp. 1- 11.
16. Langford ZL, Kumar J, Hoffman FM, "Convolutional Neural Network Approach for mapping Arctic vegetation using multi-sensor remote sensing fusion", International Conference on Data Mining Workshops, IEEE, pp. 322-331. 29
17. Abdulridha J, Ehsani R, De Castro A, "Detection and differentiation between Laurel Wilt disease, phytophthora disease and salinity damage using a hyperspectral sensing technique", International Journal on Agriculture 2016, Vol. 6, Issue 4, pp. 56-60.
18. Moshe D, Bravo C, West J, Wahlen S, McCartney A, Ramon H, "Automatic detection of Yellow Rust in wheat using reflectance measurements and neural networks", Computer Electron Agricultural Journal, 2004, Vol. 44, Issue 3, pp. 173-188.
18. Dean, Jeff Monga, Rajat. et al; "TensorFlow: Large-scale machine learning on heterogeneous systems", TensorFlow.org.Google Research, Retrieved November 10, 2015.
19. " Prediction of Dengue Disease Cases by ML Techniques", International Journal of Data Science and Machine Learning (IJDSML), ISSN : 2692-5141, Vol-1 Issue-1, Sep 2020, Page No: 1-6.
20. "An Approach For Detecting Phishing Attacks Using Machine Learning Techniques", Journal of Critical Reviews (JCR), ISSN : 2394-5125, Vol-7 Issue-18, Jun 2020, Page No: 321-324.
21. "Disease Prediction and Diagnosis Implementing Fuzzy Neural Classifier based on IoT and Cloud", International Journal of Advanced Science and Technology (IJAST), ISSN : 2005-4238, Vol-29 Issue-5, May 2020, Page No: 737-745.
22. "Research of Feature Selection Methods to Predict Breast Cancer", International Journal of Recent Technology and Engineering(IJRTE), ISSN : 2277-3878, Vol-8 Issue-2s11, Sep 2019, Page No: 2353-2355.
23. "A Novel Approach Towards Smart Agriculture Using IoT" International Journal of Research, ISSN : 2236-6124, Vol-8 Issue-4, Apr 2019, Page No: 273-278.