



Plant Leaf Disease Recognition Using Deep Learning Approach

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PLANT LEAF DISEASE RECOGNITION USING DEEP LEARNING APPROACH

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ABSTRACT: Sharp developing system using crucial establishment is an inventive advancement that improves the quality and measure of cultivating creation in the country. Plant leaf disease has for a long while been one of the critical perils to sustenance security since it definitely diminishes the reap yield and deals its quality. Correct and correct examination of disorders has been a tremendous test and he progressing propels in PC vision made possible by significant learning has arranged for camera-helped disease assurance for plant leaf. It delineated the inventive course of action that gives capable ailment ID and significant learning with convolutional neural frameworks has gained unbelievable ground in the request for various plant leaf diseases. A grouping of neuron-wise and layer-wise recognition methods were applied using a CNN, arranged with a straightforwardly open plant disease given picture dataset. Right now, saw that neural frameworks can get the tints and surfaces of wounds express to specific afflictions upon examination, which resembles human dynamic.

KEYWORDS: Disease Detection, Deep learning, Tensorflow

I. INTRODUCTION

Significant learning is a piece of AI which is completely established on counterfeit neural frameworks, as neural framework will mimic the human cerebrum so significant learning is moreover a kind of duplicate of human psyche. It's on exposure nowadays because past we didn't have that much taking care of intensity and a lot of data. A legitimate significance of significant learning is-neurons Deep learning is a particular kind of AI that achieves phenomenal power and versatility by making sense of how to address the world as settled levels of leadership of thoughts, with each thought described comparing to simpler thoughts, and continuously extraordinary depictions enlisted similar to less calculated ones. In human cerebrum around 100 billion neurons all together this is a picture of an individual

neuron and each neuron is related through a large number of their neighbors. The request here is the methods by which it recreates these neurons in a PC. Right now, makes a phony structure called a phony neural net where we have centers or neurons. It has a couple of neurons for input worth and some for yield regard and in, there may be heaps of neurons interconnected in the disguised layer.

II. RELATED WORK

The plant leaves are affected by bacterial, fungal and viral diseases which include leaf rust, powdery mildew, bacterial blight, Downey mildew, brown spot etc. Fig. 1 illustrates the classification of the bacterial, fungal and viral diseases. J. D. Pujari, R. Yakkundimath, and A. S. Byadgi applied

Artificial Neural Network, Probabilistic Neural Network, and Support Vector Machine for vegetable crops, commercial crops and cereal crops respectively for disease detection [1-4]. Balasubramanian Vijayalakshmi and Vasudev Mohan applied Fuzzy-Relevance Vector Machine classifiers in which the inputs like training features and the labels are used for leaf disease detection [5-7]. X. Wang, M. Zhang, J. Zhu and S. Geng foretold *Phytophthora infestans* disease diagnosis on tomatoes by using Artificial Neural Networks [8-10]. Dong Pixia and Wang Xiangdong proposed an approach called Minimum Distance Classifier for recognizing cucumber leaf disease [11-15]. S. Arivazhagan, R. Newlin Shebiah, S. Ananthi and S. Vishnu Varthini proposed an algorithm for classifying diseases of plants including jackfruit, tomato, etc. by using Support Vector Machine classifier [16-18]. Table I demonstrates the comparative study of disease detection in different plant leaves.

III. LITERATURE SURVEY

TITLE: (Differential) Co-Expression Analysis of Gene Expression: A Survey of Best Practices.

AUTHOR: Hussain A. Chowdhury, Dhruva K. Bhattacharyya, Jugal K. Kalita

YEAR: 2019

It presented an overview of best practices in the analysis of co-expression, expression networks, differential networking, and differential connectivity that can be discovered in microarrays and RNA-seq data, and shed some light on the analysis of scRNA-seq data as well. It has discussed co-expression analysis for RNA-seq along with a comparison of analysis of co-expression networks in microarrays vs. RNAseq. We talked about differential co-articulation and differential systems administration alongside an examination of differential articulation, differential systems

administration and differential availability. It talked about natural understanding and useful investigation to separate organic data from a lot of given qualities that are distinguished through examination. It displayed a few suggestions and rules for the examiner. Investigation of quality articulation information is generally utilized in transcriptomic concentrates to comprehend elements of atoms inside a cell and collaborations among particles. Differential co-articulation investigation considers ailments and phenotypic varieties by discovering modules of qualities whose co-articulation designs change across conditions. It review the best practices in gene expression data analysis in terms of analysis of (differential) co-expression, co-expression network, differential networking, and differential connectivity considering both microarray and RNA-seq data along with comparisons. It highlights hurdles in RNA-seq data analysis using methods developed for microarrays. It includes discussion of necessary tools for gene expression analysis. In addition, it shed light on scRNA-seq data analysis by including preprocessing and scRNA-seq in co-expression analysis along with useful tools specific to scRNA-seq. To get insights, biological interpretation and functional profiling is included. Finally, it provides guidelines for the analyst, along with research issues and challenges which should be addressed.

TITLE: Image of plant disease segmentation model based on pulse coupled neural Network with shuffle frog leap algorithm

AUTHOR: Xiaoyan Guo, MingZhang, Yongqiang Dai

YEAR: 2018

A novel image segmentation model SFLA-PCNN for plant diseases based on hybrid frog-hopping algorithm is proposed.

Using the weighted sum of cross entropy and image segmentation compactness as the fitness function of SFLA, the image of potato late blight disease is taken as a trial segmentation image to find the optimal configuration parameters of PCNN neural. Picture division is a key advance in include extraction and illness acknowledgment of plant ailments pictures. To stay away from the subjectivity of utilizing customary PCNN (beat coupled neural system) to section plant ailment picture, another picture division model (SFLA-PCNN) is proposed get the parameters design of PCNN. The weighted whole of cross entropy and conservativeness level of picture division is picked as wellness capacity of rearranged frog jump calculation to advance the parameters PCNN, which could improve the presentation of PCNN. After multiple times neighborhood emphasis and multiple times worldwide cycle, we get the best parameter arrange. The broad tests demonstrate that SFLA-PCNN model could be utilized to remove the injury from the foundation viably, which could give an establishment to following illness analyze.

TITLE: A Smart Phone Image Processing Application for Plant Disease Diagnosis network with shuffle frog leap algorithm
AUTHOR: Nikos Petrellis
YEAR: 2017

An advanced mobile phone application for plant sickness acknowledgment was exhibited. It depends on picture preparing that breaks down the shading highlights of the spots in plant parts. It was assessed on grape ailments with a precision that surpasses 90% utilizing a little preparing set. Albeit proficient agribusiness engineers are liable for the acknowledgment of plant sicknesses, smart frameworks can be utilized for their finding in beginning periods. The master frameworks that have been proposed in the writing for this reason for

existing are frequently founded on realities portrayed by the client or picture preparing of plant photographs in noticeable, infrared, light and so on. The acknowledgment of an infection can regularly be founded on side effects like injuries or spots in different pieces of a plant. The shading, territory and the quantity of these spots can decide, all things considered, the illness that has embarrassed a plant. Greater expense sub-atomic investigations and tests can follow if important. A Windows Phone application is portrayed here equipped for perceiving vineyard maladies through photographs of the leaves with exactness higher than 90%. This application can without much of a stretch be reached out for various plant maladies and diverse advanced cell stages.

TITLE: Plant Diseases Recognition for Smart Farming Using bModel-based Statistical Features
AUTHOR: Chit Su Hlaing, Sai Maung Maung Zaw
YEAR: 2017

It has demonstrated the benefits of GP circulation model for SIFT descriptor and effectively applied in plant illness characterization. Besides, it proposed include accomplishes a decent tradeoff among execution and grouping precision. In spite of the fact that it proposed highlight can effectively show the SIFT include and applied in plant ailments acknowledgment, it have to attempt to improve our proposed include by considering and participation with other picture preparing strategies. It examine is to distinguish and characterize the plant ailment in horticultural space, by actualizing picture preparing strategies. The info pictures are taken by different mobile cameras. The Scale-invariant feature transform (SIFT) features used as texture feature and it is invariant to scaling, rotation, noise and illumination. But the exact mathematical

model of SIFT texture descriptor is too complex and take high computing time in training and classification. The model-based statistical features are calculated from SIFT descriptor to represent the features of an image in a small number of dimensions. It derive texture information probability density function called Generalized Pareto Distributions from SIFT texture feature. It proposed feature is to reduce computational cost of mobile devices. In our experiment, 10-Fold cross validation with SVM classifiers are applied to show that our experiment has no data bias and exclude theoretically derived values.

TITLE: Pepper Cutting UGV and Disease Detection using Image Processing

AUTHOR: Rakshitha.N, Rekha.H.S, Sandhya.S, Sandhya.V and S. Sowndeswari.

YEAR: 2017

It proposed system a simple UGV is designed and implemented. Pepper cutting is a serious issue as it grows to greater heights and is most likely to get affected by the diseases. To overcome these limitations “pepper cutting UGV and disease detection using image processing” is implemented which in turn reduces the human efforts. The pepper plucking and disease detection can be done with the help of image processing technique. Agriculture is the backbone of India. Over 77% of the Indian population depends on agriculture. The crop yield is reduced when the plants are infected by diseases. It disturbs the economic conditions of the farmers which in turn reduces the productivity of the nation. The proposed system focuses on designing an UGV (Unmanned Ground Vehicle) for the purpose of cutting pepper fruit, disease detection and also providing solution for the disease by sprinkling the required pesticides. The paper discusses the extraction of mainly the pepper plant based on the color recognition

techniques and also the methods used for the detection of plant diseases using the images of leaves and providing solution by opting sprinkling mechanism.

V. COMPARISON

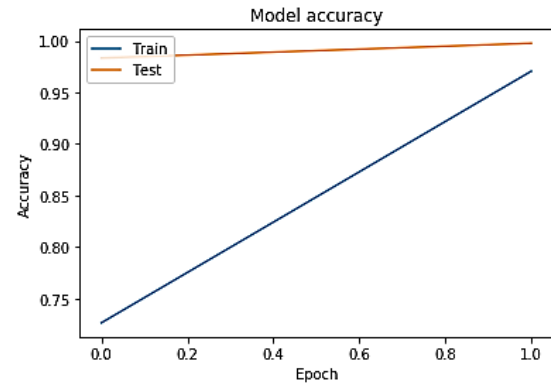


Fig.1: CNN model trained dataset accuracy

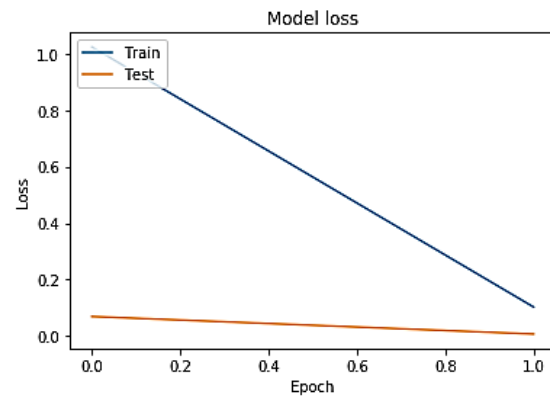


Fig.2: CNN model trained dataset loss values

VI. EXISTING SYSTEM

Plants are viewed as significant as they are the wellspring of vitality supply to humanity. Plant infections can influence the leaf whenever among planting and collecting which prompts colossal misfortune on the creation of yield and practical estimation of market. In this

manner, leaf ailment identification assumes a crucial job in farming field. Nonetheless, it requires gigantic labor, additionally preparing time and broad information about plant maladies. Henceforth, AI is applied to recognize illnesses in plant leaves as it breaks down the information from various angles, and arranges it into one of the predefined set of classes. The morphological highlights and properties like shading, force and measurements of the plant leaves are mulled over for characterization and the different kinds of plant illnesses and distinctive arrangement procedures in AI that are utilized for distinguishing infections in various plant leaves.

VII. PROPOSED SYSTEM

We wanted to structure profound learning method so an individual with lesser aptitude in programming ought to likewise have the option to utilize it no problem at all. It proposed framework to foreseeing leaf ailments. It clarifies about the trial examination of our technique. Tests of 38 pictures are gathered that included distinctive plant ailments like Apple, Tomato, Grape and Healthy Leaves. Distinctive number of pictures is gathered for every sickness that was ordered into database pictures and information pictures. The essential qualities of the picture are depended upon the shape and surface arranged highlights. The example screen captures shows the plant leaf sickness discovery utilizing shading based division model.

VIII. MODULE DESCRIPTION

1. TRAIN CNN (MODULE01)
2. WORKING PROCESS OF LAYERS IN CNN MODEL (MODULE02)
3. PLANT LEAF DISEASE IDENTIFICATION (MODULE03)

TRAIN CNN (MODULE01)

The dataset consists of concerning 84K RGB pictures of healthy and unhealthy crop leaves are collected that is categorized into thirty eight completely different categories. The overall dataset is split into 80/20 quantitative relation of coaching and validation set protective the directory structure.

(<https://www.kaggle.com/vipooooool/new-plant-diseases-dataset>)

A dataset consisting of concerning thirty eight completely different plant leaf diseases is obtained, out of that any image may be used as a check image for the software system.

We have to import our information set mistreatment keras preprocessing image information generator perform additionally we tend to produce size, rescale, range, zoom range, horizontal flip. Then we tend to import our image dataset from folder through the info generator perform. Here we tend to set train, test, and validation additionally we tend to set target size, batch size and class-mode from this perform we've to coach.

To prepare our dataset utilizing classifier and work generator work likewise we tend to create making ready steps per ages at that time complete range of ages, approval info and approval steps utilizing this info we are able to prepare our dataset.

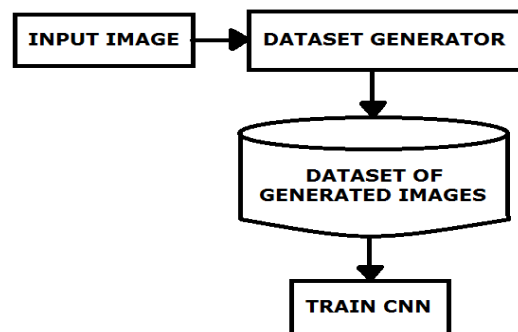


Fig.3: Train CNN

WORKING PROCESS OF LAYERS IN CNN MODEL (MODULE02)

A Convolutional Neural Network (ConvNet/CNN) is a Deep Learning algorithm which can take in an input image, assign importance (learnable weights and biases) to various aspects/objects in the image and be able to differentiate one from the other. The pre-processing required in a ConvNet is much lower as compared to other classification algorithms. While in primitive methods filters are hand-engineered, with enough training, ConvNets have the ability to learn these filters/characteristics. The architecture of a ConvNet is analogous to that of the connectivity pattern of Neurons in the Human Brain and was inspired by the organization of the Visual Cortex. Individual neurons respond to stimuli only in a restricted region of the visual field known as the Receptive Field. Their network consists of four layers with 1,024 input units, 256 units in the first hidden layer, eight units in the second hidden layer, and two output units.

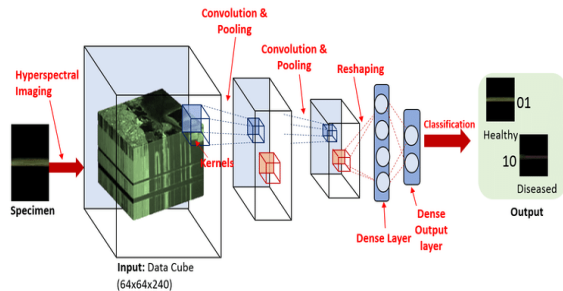


Fig.4: Working Process of Layers in CNN Model

PLANT LEAF DISEASE IDENTIFICATION (MODULE03)

We give input picture utilizing keras preprocessing bundle. That input image changed over into exhibit esteem utilizing pad and picture to cluster work bundle. We have just arranged ailment of leaf in our dataset. It orders what are the plant sickness

leaves. At that point we need to foresee our leaf maladies utilizing anticipate work.

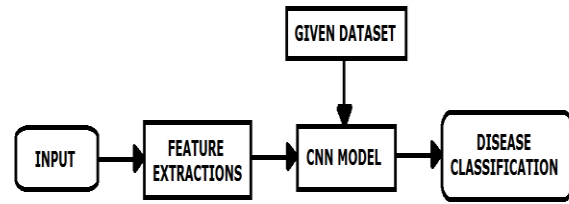


Fig.5: Plant Leaf Disease Identification

VII. SYSTEM ARCHITECTURE

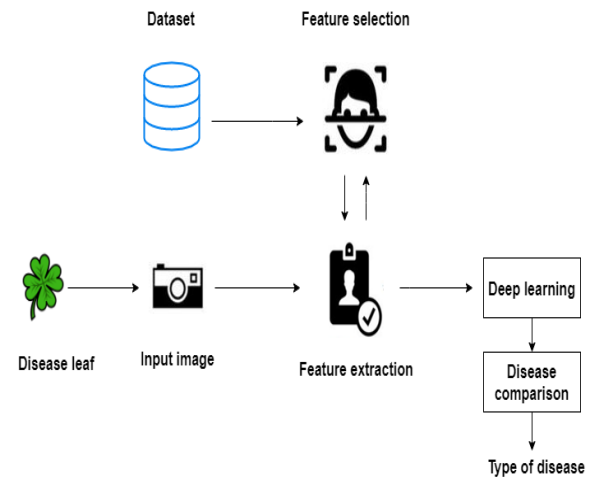


Fig.6: System Architecture

System configuration is the hypothetical model that portrays the structure, lead, and more points of view on a structure. A designing delineation is a traditional depiction and depiction of a system, created to such an extent that supports contemplating the structures and practices of the structure. A structure configuration can contain system parts and the sub-systems made, that will coordinate to execute the general system. There have been attempts to formalize lingos to delineate structure plan; with everything taken into account these are called building depiction tongues.

IX. FUTURE ENHANCEMENT

Agricultural office needs to mechanize the distinctive the yield crops from capability process. To modernize this method by show the desire achieves web application or work zone application. To improve the work to execute in Artificial Intelligence condition.

X. CONCLUSION

It focused how picture from given dataset (arranged dataset) in field and past educational assortment used foresee the case of plant afflictions using CNN model. This brings a part of the going with bits of information about plant leaf contamination estimate. As most extraordinary sorts of plant leaves will be made sure about under this system, farmer may locate a decent pace the leaf which may never have been created and runs through all possible plant leaves, it enables the farmer in unique of which to gather to create. In like manner, this structure mulls over the past formation of data which will empower the farmer to get understanding into the intrigue and the cost of various plants in feature.

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