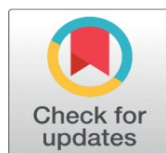
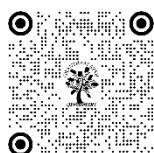


MACHINE LEARNING BASED PADDY DISEASE PREDICTION AND CLASSIFICATION USING CONVOLUTIONAL NEURAL NETWORK

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DOI

[10.29121/shodhkosh.v5.i5.2024.3277](https://doi.org/10.29121/shodhkosh.v5.i5.2024.3277)

Funding: This research received no specific grant from any funding agency in the public, commercial, or not-for-profit sectors.

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ABSTRACT

The essential wellspring of nourishment for everybody is horticulture. Broad review shows that farming is essentially affected by the advancement of ailments in rice fields. Ranchers are confronted with the difficult undertaking of unequivocally distinguishing contamination in their corn crops during the foliar illness stage. Hence, to foresee illnesses, programmed sickness discovery calculations should focus more on early signs of paddy infections. The sorts of harvests that can be developed and their yields are fundamentally affected by these sicknesses, which first assault the leaves prior to spreading all through the whole rice field. Subsequently, to successfully control the spread of disease and backing solid plant development, exact classification and early distinguishing proof of paddy sicknesses are fundamental. This article proposes a convolutional neural network (CNN) model for paddy sickness order with an end goal to defeat these hardships. The proposed approach searches for sores on the leaves that could be the consequence of an illness utilizing picture handling calculations. Through preprocessing, CNN's component testing precision with classifier is better than that of other current models. Utilizing genuine Inaba photographs, the viability of the proposed consideration-based preprocessing was additionally affirmed. The CNN model has a higher characterization exactness when contrasted with other exchange learning calculations.

Keywords: Paddy Disease, Corn Crop Infections, CNN, Image-Processing. Convolutional Neural Network (CNN) Model

1. INTRODUCTION

India's primary source of income is agriculture. Almost 70% of people in India are employed in agriculture. Approximately 58% of those living in rural areas rely on agriculture as their primary source of income. An essential food in India is rice. In Asia, the rice disease wipes out 10–15% of the crop. Plant diseases are caused by bacteria and fungi. Sheath blight, brown spot, leaf burn, and leaf blast are among the diseases that affect rice.

Over half of the world's population depends on rice, making it one of the most essential food crops. Since 2015, India has emerged as the world's second-largest producer and exporter of rice. India's economy is so heavily reliant on the production of rice. However, a number of illnesses that affect crops annually cause yield losses for farmers. In certain instances, farmers might not be able to detect the illness in time or might have trouble treating it, which would cause

them to recover from the disease too slowly and with considerable production losses. Identification, detection, and prompt application of remedies are essential components of reducing the detrimental effects of illness.

Farmers had to use reference books or their own knowledge to identify diseases, which made it extremely difficult to choose the best pesticide quickly. Autonomous systems for rice disease classification can be developed thanks to advances in data mining and image processing. To identify diseases, regular photos of various rice plant components are obtained and processed through a number of processes, including feature extraction, image segmentation, image preprocessing, image acquisition, and disease classification.

To identify diseases, ranchers needed to depend on manuals or draw on their own insight, which expected an extremely tedious cycle to choose the right pesticide as soon as possible. Current picture handling and information mining improvements permit the advancement of independent frameworks for characterizing rice sicknesses. Normal pictures of various pieces of rice plants are procured and handled utilizing a progression of steps, for example, picture securing, picture preprocessing, picture division, highlight extraction, and infection characterization to perceive illnesses.

1.1. PLANT DISEASES

Plant diseases that happen at various pieces of the plant. In this paper, we have confined our work inside the normal illnesses in India. This segment will give some thought regarding which picture handling procedures are reasonable and what sort of highlights is pertinent to precisely identify infection. To work on the sickness recognition framework, we have ordered the illnesses into four classes in view of the area of the contamination in the plant body. Prior to delving into subtleties of the illnesses, we should be know all about the area of the contaminated pieces of the adult. Plant for more exact identification. The plant comprises of grains/panicles, leaf, sheath, stem (counting hub, neck), and root. At first, need to isolate four nonexclusive sorts of illnesses happen in separate pieces of the plant.

Grains/Panicle Misleading Filth: The tainted plant produces smooth spores from grain. Development of smooth spores encases botanical parts. Smooth and minimal leveled yellowish spores demonstrate that the spores are immature which is covered by film. Development of spores brings about broken layer. Mature spores are orange and become yellowish-green or greenish-dark in variety. Grains/Panicle — Grain Staining: It happens at the hour of panicle commencement stage. As of now, typical grain variety changes to earthy white tone.

Disease on plant prompts the huge decrease in both the quality and amount of agrarian items. The investigations of plant sickness allude to the investigations of outwardly perceptible examples on the plants. Observing of wellbeing and illness on plant assumes a significant part in effective development of yields in the ranch. In early days, the observing and examination of plant illnesses were done physically by the ability individual in that field. This requires colossal measure of work and furthermore requires over the top handling time. The picture handling strategies can be utilized in the plant illness identification. In the greater part of the cases sickness side effects are seen on the leaves, stem and organic product. The plant leaf for the discovery of infection is viewed as which shows the illness side effects. The prologue to picture handling strategy utilized for plant illness discovery.



Fig.1. plant diseases detection

The resized pictures conveying most ailing parts, physically chose for division. In this examination, six classes of plant pictures with 30 pictures with every classification had thought of. The five classes had a place with infected plant, and one classification was a cleaned picture. The considered tainted plants were the Earthy colored spot, Panicle impact,

Leaf impact, Stem Drill and Sheath scourge. The preview of the pre-handled panicle impact contaminated 30 pictures is portrayed.

2. RELATED WORK

P. Jiang et al (2019): Alternaria leaf spot, Earthy colored spot, Mosaic, Dim spot, and Rust are five normal sorts of apple leaf infections that seriously influence apple yield. Notwithstanding, the current examination misses the mark on exact and quick finder of apple illnesses for guaranteeing the solid advancement of the apple business. A profound learning approach that depends on better Repetitive brain organizations (RNNs) for the constant recognition of apple leaf infections. In this paper, the apple leaf sickness dataset (ALDD), which is made out of research facility pictures and complex pictures under genuine field conditions, is first developed through information expansion and picture explanation advancements. In light of this, another apple leaf sickness recognition model that utilizes profound RNNs is proposed by presenting the Google Net Commencement construction and Rainbow link. At last, under the hold-out testing dataset, utilizing a dataset of 26,377 pictures of sick apple leaves, the proposed INAR-SSD (SSD with Origin module and Rainbow connection) model is prepared to identify these five normal apple leaf illnesses.

Q. Zeng, et al (2020): Deep Learning models have been utilized to arrange various types of plant sicknesses, yet very little work has been finished for illness seriousness discovery. In any case, it is more critical to dominate the severities of plant illnesses precisely and ideal, as it assists with pursuing successful choices to shield the plants from being additionally contaminated and diminish monetary misfortune. In this paper, in view of the tainted leaf pictures got from Plant Town and group, a dataset with 5,406 citrus leaf pictures contaminated by HLB. Then, at that point, six various types of well-known models were prepared to play out the seriousness location of citrus HLB with the objective to find which kinds of models are more reasonable to distinguish HLB seriousness with a similar preparation situation. The trial results show that the Inception_v3 model with epochs=60 can accomplish higher precision than that of different models for seriousness recognition with an exactness of 74.38% because of its exceptionally computational proficiency and modest number of boundaries.

U. P. Singh, et al (2019): Contagious sicknesses not just impact the monetary significance of the plants and its items yet additionally lessen their natural conspicuousness. Mango tree, explicitly the foods grown from the ground leaves are exceptionally impacted by the contagious sickness named as Anthracnose. The fundamental is to foster a fitting and powerful strategy for conclusion of the sickness and its side effects, thusly embracing a reasonable framework for an early and practical arrangement of this issue. Throughout the course of recent years, because of their better exhibition capacity concerning calculation and precision, PC vision, and profound learning procedures have acquired notoriety in grouped contagious illnesses characterization. Hence, for this paper, a multilayer Recurrent neural network (MRNN) is proposed for the order of the Mango leaves tainted by the Anthracnose contagious illness.

G. Yang et al (2020): Artificial recognition of tomato sicknesses is in many cases tedious, arduous and abstract. For tomato infection pictures, it is hard to track down little discriminative highlights between various tomato illnesses, which can carry difficulties to fine-grained visual arrangement of tomato leaf-based pictures. A model, which comprises of 3 organizations, including an Area organization, a Criticism organization, and an Order organization, named LFC-Net. Simultaneously, a self-management system is proposed in the model, which can really recognize useful districts of tomato picture without the requirement for manual explanation, for example, bouncing boxes/parts. In view of the thought of the consistency between classification of the picture and in development of the picture, we plan a clever preparation worldview. The Area organization of the model initially distinguishes enlightening locales in the tomato picture, and improves cycles under the direction of the Criticism organization. Then, at that point, the Arrangement network utilizes instructive locales proposed by the Area organization and the full picture of the tomato for characterization. Our model can be viewed as a multi-network cooperation, and organizations can advance together.

T. N. Pham et al (2020): Plant illness, particularly crop plants, is a significant danger to worldwide food security since numerous sicknesses straightforwardly influence the nature of the natural products, grains, etc, prompting a lessening in horticultural efficiency. Ranchers need to notice and decide if a leaf was contaminated by unaided eyes. This interaction is untrustworthy, conflicting, and blunder inclined. A few deals with profound learning procedures for identifying leaf infections had been proposed. The majority of them fabricated their models in light of restricted goal pictures utilizing Repetitive Brain Organizations (RNNs). In this exploration, we target recognizing early sickness on plant leaves with little illness masses, which must be identified with higher goal pictures, by a Fake Brain Organization

(ANN) approach. After a pre-handling step utilizing a difference improvement strategy, every one of the plagued masses are fragmented for the entire dataset. A rundown of a few estimations based highlights that addresses the masses are picked and afterward chose in view of their effects on the model's presentation utilizing a covering based include determination calculation, which is fabricated in light of a mixture metaheuristic.

X. Yang et al (2020): an occasion division and characterization strategy for plant leaf pictures in view of IFPN SNMS CFFI-Veil R-RNN (ISC-MR RNN) and ACPSSOSVM-Double Channels Intermittent Brain Organization (APS-DC RNN). To acquire the forefront of plant leaf pictures, the sidelong association construction of the component map pyramid in ISC-MR RNN wires the element guides of various profundities, with the goal that the organization learns more nitty gritty highlights. Then, at that point, the Delicate Non-Most extreme Concealment Calculation is utilized to further develop the discovery execution of covering objects. Then, the pooling technique for incorporating the persistent capability can decrease the accuracy misfortune during the arrangement of the planning between the element map and the first picture. At long last, by building a cover channel layer, complex foundations are concealed. To recognize the similitude between plant leaf pictures, APS-DC RNN is utilized to order the frontal area pictures. In this cycle, the Help Vector Machine is utilized to supplant delicate max and afterward a Versatile Turbulent Molecule Multitude Calculation is utilized to streamline it.

T. Akiyama et al (2019): The capacity to recognize plant types is significant while directing vegetation overviews. This capacity requires agents experience. A portable application utilizing Recurrent neural networks (RNN s) that will assist fledglings with distinguishing plant species. Look at three RNN models, VGG19, Versatile Net, and MobileNetV2. Our plant distinguishing proof application utilizing MobileNetV2 shows a typical F1 score of 0.992, demonstrating its superior presentation and reasonableness. The carried out framework shows a viable presentation of 338.1 ms per picture on a tablet-type gadget.

S. Ghosal et al (2020): plant is one of the major developed crops in India which is impacted by different illnesses at different phases of its development. It is truly challenging for the ranchers to physically recognize these illnesses precisely with their restricted information. Late improvements in Profound Learning show that Programmed Picture Acknowledgment frameworks utilizing Recurrent Neural Network (RNN) models can be extremely valuable in such issues. Since plant leaf sickness picture dataset isn't effectively accessible, we have made our own dataset which is little in size consequently we have utilized Move Figuring out how to foster our profound learning model. The proposed RNN design depends on VGG-16 and is prepared and tried on the dataset gathered from rice fields and the web and the precision of the proposed model is 92.46%.

G. Singh, et al (2020): The trees are fundamentally seen by techniques for their leaves. There are one kinds of timber fostered all through the world, some are basic money yield and some are used in prescription. Plant care is outstandingly fundamental in agribusiness for the association of plant species however botanists can use this utility for helpful purposes. Leaf of select vegetation has one of the sort credits that can be used to arrange them. Straightforward and computationally condition technique for plant ID using modernized photograph taking care of and figuring contraption vision innovation. This model basically involves three stages which consolidate pre-getting ready of leaf data, ascribes extraction from colossal datasets and the last one is structure. Pre-planning is the strategy of updating estimations pix before computational taking care of. The limit extraction part decides real factors subject to the shade and construction of the leaf picture. These points are used as commitments to the classifier for condition cheerful request and the outcomes were attempted and strangely the usage of Artificial Neural Network (ANN) and Euclidean (KNN) classifier.

S. S. Hari et al (2019): The powerful order of plant infections. By and large, people can distinguish the plants that are impacted by specific sicknesses, yet aside from our visual perception, it is difficult to recognize. Without giving the right treatment and prompt activities, the whole development land can transform into a sickness impacted region, else all plants which are a neighbor to each other can get impacted through spreading. In this way, to recognize the plant illnesses ahead of time and to identify the sicknesses with the assistance of current PC innovation, we proposed a model for the proficient distinctive plant illnesses. The dataset utilized here comprises of a few assortments of plants of both impacted and sound, and this large number of pictures are gathered from different uninhibitedly accessible sources and physically. Another RNN model was prepared and tried. At long last, this proposed model has given excellent exactness when tried in field conditions.

3. IMPLEMENTATION OF PROPOSED METHOD

To distinguishing and characterizing the plant sickness are utilized for improvement of proposed framework. Convolutional Neural Network (CNN) Characterizing the various sorts of plant sickness are arisen as exploring process in valuable horticulture field. By utilizing this mechanized technique, it can limit the quantitative and subjective misfortune in farming field. This course of limiting the misfortune can be executed by involving the computerized picture handling methods in cultivating field. This programmed recognizing procedure in picture handling makes the exact and proficiency in tracking down the sicknesses in plant. By utilizing this procedure can limit the human vision and more over manual technique finds opportunity to examine the infection and it won't be precise, However CNN exact to foresee the illnesses.

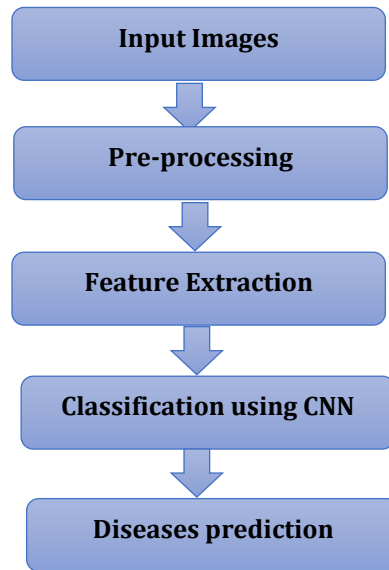


Fig. 2. Proposed Diagram

To see if the leaf is unhealthy or sound, certain means should be followed. Preprocessing Element extraction, Preparing of classifier and Grouping. Preprocessing of picture is bringing every one of the pictures size to a decreased uniform size. Then comes removing highlights of a preprocessed picture which is finished with the assistance. A component descriptor utilized for object discovery. In this element descriptor the presence of the item and the framework of the picture is depicted by its power slopes. One of the benefits of component extraction is that it works on the cells made. Any changes don't influence this. This cycle utilizing leaf illness find and trimming in the CNN method here we utilized three element descriptors minutes: Picture minutes which have the significant attributes of the picture pixels helps in portraying the items. Here help in portraying the blueprint of a specific leaf. are determined over single channel as it were. The initial steps include changing over dark scale and afterward are determined. This step gives a variety of shape descriptors. The commitment of the plant leaf dataset to foster Man-made reasoning models (AI and profound learning) will help numerous analysts and PC researchers to recognize, distinguish the species and its sicknesses and become familiar with the spice presence and properties. By delivering this dataset to the local area, we anticipate animate exploration in plants where the ongoing absence of public datasets is one of the primary obstructions for progress.

3.1. TRAINING DATASET

Another alteration of CNN troupe preparing pictures is preprocessing and arrangement by joining capabilities from cutting edge profound CNN models for leaf picture acknowledgment. We initially apply the section leaf pictures from the foundation to work on the exhibition of the acknowledgment framework. Then, we present a model methodology in view of a blend of misfortune capabilities from the CNN to sum up a capability. The joint learning multiclass model intended for leaf acknowledgment permits each organization to play out its undertaking and help out the others all the while,

where information from different prepared profound organizations is shared. This participation proposed multimodal is compelled to manage more convoluted issues instead of a straightforward characterization.

In this manner, the learn can learn a lot of rich data and further develop its speculation capacity. Moreover, a multiclass compromise procedure between two profound learning models can diminish the impact of overt repetitiveness issues in gathering classifiers. The presentation of our methodology is assessed 80% our custom Vietnamese home grown leaf species dataset, and public datasets, for example, Leaf snap, and Folio are utilized to construct experiments. The outcomes affirm that our methodology improves the leaf acknowledgment execution and outflanks the ongoing standard single organizations while having less low calculation cost.

3.2. TESTING DATASET

These plants are valuable in numerous ways like medication definition, creation of natural items, and meds to fix numerous normal illnesses and sicknesses. A conventional Indian therapeutic framework is broadly acknowledged even today. India is a rich country for being the territory for different restorative plants. Many pieces of the plants, for example, preprocessing and characterization element for the development of natural drugs. Are liked in both creating and created nations as an option in contrast to engineered sedates fundamentally in view of no aftereffects.

The component work Acknowledgment of these weeds by human sight will be drawn-out, tedious, and wrong. Utilizations of picture editing and PC vision strategies for the recognizable proof of the restorative weeds are extremely essential however many of them are under elimination according to the dataset. Consequently, the digitization of helpful restorative plants is pivotal for the preservation of biodiversity. The exhibition of our methodology is assessed 20% our custom Vietnamese home grown leaf species dataset for acknowledgment of restorative spices requires a nice size of plant leaf dataset.

3.3. CONVOLUTIONAL NEURAL NETWORK (CNN)

In this part depict our profound learning convolutional neural network (CNN) model and its preparation. We have involved the TensorFlow system for execution of the CNN model. TensorFlow purposes alleged information stream charts for mathematical calculation. Every hub of the diagram addresses numerical operation(s) and the chart edges address multi-faceted exhibits or tensors imparted between them. Along these lines, execution of an AI calculation is direct in TensorFlow. We have made a complex profound learning model with different convolutional channel on figure.3.

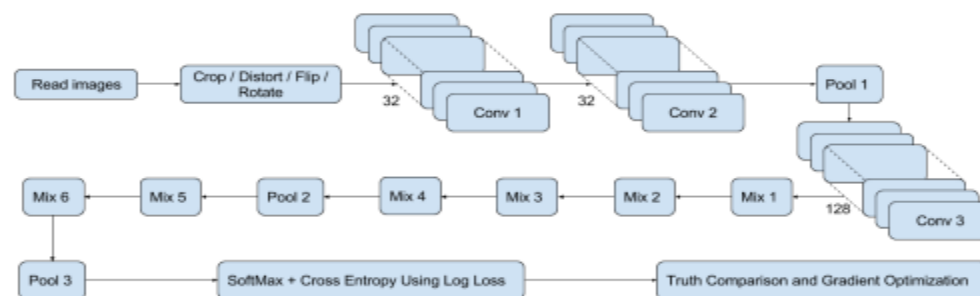


Fig. 3. Convolutional neural network (CNN)

A Convolution step includes putting a channel of a specific size at various pieces of the tensor and working out the dab item between the channel and the tensor. This single worth is taken as the result an incentive for that place of the channel. The separating stretch for arrangement of the convolution channel is called step. For instance, in Conv1, 32 convolution channels of 3x3x3 size are applied to the result tensor of the past step of size 256x256x3. By changing the step of the channels, aspects of the result can be controlled. In Conv1, a step aspect of 2x2x1 (2x2 in pixels course and 1 in variety heading) successfully parts the size of the result in pixel bearing.

Along these lines, the result of every convolution channel is a tensor of size $128 \times 128 \times 1$ and the joined result of 32 channels is a tensor of size $128 \times 128 \times 32$. Essentially, in Conv2, 32 convolution channels of size $3 \times 3 \times 32$ with step of $1 \times 1 \times 1$ are applied, bringing about a result size of $128 \times 128 \times 32$. One more piece of the model, the Pooling step, works by applying a cover onto various pieces of the picture and choosing a solitary worth from inside the veil as the result esteem at that position. In the event of Max Pooling, the most extreme worth is taken.

4. RESULT AND DISCUSSION

SAR is generally used for remote sensing purposes. Except for the military operations, it will be used for private purposes. This is, SAR image imaging, ocean current patterns for weather and climate monitoring, will help to move the pattern and volcanic movement of the glacier, and geological applications.

No. of data	LR %	RF %	CNN
10	41	55	70
20	28	22	80
30	55	40	90
40	60	58	95

Table 1 shows the accuracy performance on the table value for the percentage level of algorithm performance high level for migration learning algorithm and SAR imaging algorithm.

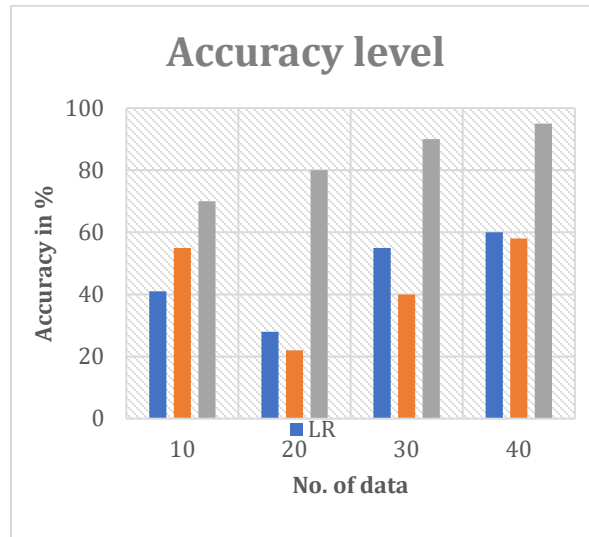


Fig.4. Accuracy Level performance

Figure.4 shows the accuracy level performance on the proposed method for migration learning algorithm and SAR imaging algorithm performance. Existing Logistic Regression (LR) 60% and then Random Forest (RF) result is 58% level. The proposed migration learning algorithm and Synthetic Aperture Radar (SAR) imaging algorithm result is 95% of the accuracy level.

No. of data	LR (ms)	RF(ms)	CNN
10	55	44	30
20	45	30	25
30	20	35	20
40	50	40	15

Table 2 shows the time complexity performance on the table value for the algorithm performance on the time complexity performance level.

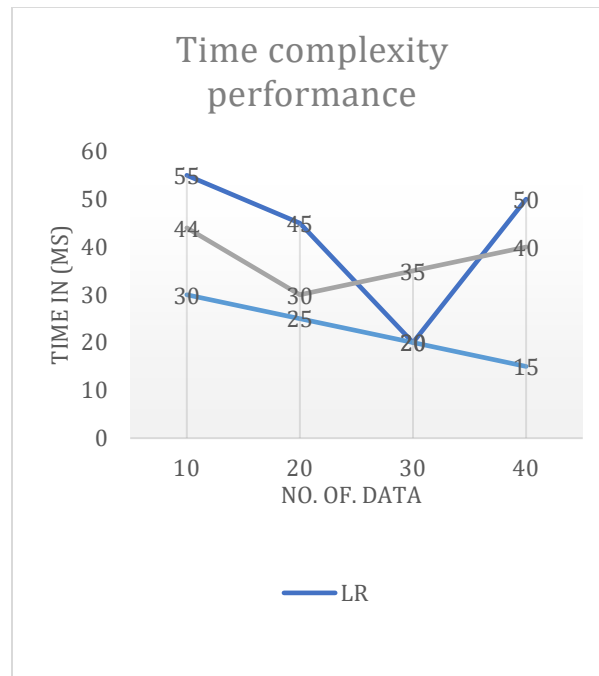


Fig. 5. performance of time complexity

Figure 4 shows the timing performance of the proposed method for migration learning algorithm and SAR imaging algorithm performance. Existing Logistics Regression (LR) 50% and then Random Forest (RF) result is 40% level. The proposed migration learning algorithm and Synthetic Aperture Radar (SAR) imaging algorithm result is 15% of the time complexity level.

5. CONCLUSION

The precisely location and order of the plant illness is vital for the effective development of yield and this should be possible utilizing picture handling. In the strategies to fragment the illness part of the plant. A few Component extraction and grouping strategies to extricate the elements of contaminated leaf and the order of plant illnesses. The utilization of CNN strategies for characterization of sickness in plants, for example, self-putting together element map, back proliferation calculation and so on can be productively utilized. From these strategies, it can precisely distinguish and group different plant sicknesses utilizing picture handling procedures. The framework was prepared by utilizing Beast force matcher CNN classifier. It precisely recognizes the infected spots present if any and arranged the sort of the sickness being impacted and fix ideas were given against distinguished illness. The framework was assessed utilizing accuracy and review technique and viewed as exact. The outcomes got can help the ranchers in successful dynamic which can productively safeguard their compost crops from significant harm. In future work as better datasets become accessible later on, pre-handling of pictures before model seed picture client to preparing in CNN can demonstrate priceless to accomplish high true execution.

CONFLICT OF INTERESTS

None.

ACKNOWLEDGMENTS

None.

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