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Plant Leaf Disease Identification using Machine Learning Techniques: A Review

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ABSTRACT: AI technologies help farmers to analyze land/soil/health of crop etc and save time and allow farmers to grow right crop in each season that has best yield. Vertical cropping can reduce water usage, make efficient land usage, can be cultivated in urban areas in buildings. It can reduce the problems with labor unavailability. It allows prediction of next year crop seasons, weather, climate, rainfall etc. AI based predictions enable suggesting appropriate pesticides, crops, place at right time before large scale incidence of disease. This paper presents review of machine learning techniques for plant leaf disease identification.

KEYWORDS: AL, Plant, Disease, Machine Learning, Crop.

I. INTRODUCTION

As corona virus spread around the world, many significant grain-delivering nations have taken on measures to confine their grain trades; food security hosts excited incredible worry from different gatherings. Instructions to further develop grain creation has gotten perhaps the main issues confronting all nations. Nonetheless, crop illnesses are a troublesome issue for some ranchers so ace the seriousness of harvest sicknesses ideal and precisely to assist with staffing take further mediation measures to limit plants being additionally contaminated.

Artificial Intelligence (AI) can be applied cross disciplinary and it can likewise bring a change in perspective by the way we see cultivating today. AI-controlled arrangements won't just empower farmers to accomplish more with less, it will likewise work on quality and guarantee quicker go-to-advertise for crops. The present innovation progression in Artificial Intelligence, Enormous Information, IoT are turning into the significant drivers for giving the Computerized IT arrangement practically in every one of the fields and business areas. Thus, it is proposed to utilize Advanced arrangement aided with Artificial intelligence to inspire the living space of the stomped on farmer local area while giving yet another chance to business and business visionaries by empowering keen ranch as an assistance.

Growth driven by Internet of Things (IOT):

IoT advancements permit relationships of organized and unstructured information to give bits of knowledge into food creation. Tremendous volumes of information get created each day in both organized and unstructured arrangement. These identify with information on authentic climate design, soil reports, new exploration, rainfall, bother invasion, pictures from Robots and cameras, etc. Psychological IoT arrangements can detect this information and give solid bits of knowledge to further develop yield. Vicinity Detecting and Far off Detecting are two advancements which are fundamentally utilized for insightful information combination. One use instance of this high-goal information is Soil Trying. While distant detecting requires sensors to be incorporated into airborne or satellite frameworks, closeness detecting requires sensors in touch with soil or at an extremely short proximity.



Figure 1: Sample of plant leaf disease

This aides in soil portrayal dependent on the dirt beneath the surface in a specific spot. Equipment arrangements like Robot (pertaining to corns) are now pairing information gathering programming with mechanical technology to set up the best compost for developing corns notwithstanding different exercises to amplify yield. The IoT empowered sensors should be introduced in the field at the recommended areas. These sensors are the transducers that gather the information on climatic condition, soil dampness and fruitfulness, root and shoot development; profused leaves development, photograph period observing, flower and seed setting, grain/organic product bearing, bother and expires as basic development factors indications and collect availability.

The IoT gadget incorporates the transducer that tests the different boundaries of climate and crop referenced previously. It tends to be mounted on secured smaller than expected board with Wi-Fi gadget, microcontroller, minimal expense VGA picture sensor, scaled down battery controlled with miniature sun oriented board. The information can be gathered at required time stretches either by introducing Wi-Fi dynamic problem area towers as needed for whole field inclusion. Then again, drones with dynamic Wi-Fi problem area can likewise be utilized to output and gather information from IoT gadgets just as to catch raised movie of the whole field.

Information driven cultivating: By investigating and connecting data about climate, sorts of seeds, soil quality and likelihood of diseases, chronicled information, commercial center patterns, and costs, farmers will settle on more educated choices.

Image-based knowledge age: Exactness cultivating is perhaps the most talked about regions in cultivating today. Robot based pictures can help in inside and out field investigation, crop checking, examining of fields, etc. PC vision innovation, IOT and robot information can be joined to guarantee fast activities by farmers. Feeds from drone picture information can produce cautions continuously to speed up accuracy cultivating. Organizations like Aeronautical trionics have carried out IBM Watson IoT Stage and the Visual Acknowledgment APIs in business drones for picture examination progressively. Given beneath are a few regions where PC vision innovation can be put to utilize:

Disease discovery: Pre-handling of picture guarantee the leaf pictures are fragmented into regions like foundation, non-diseased part and diseased part. The diseased part is then cropped and ships off far off labs for additional conclusion. It additionally helps in bother recognizable proof, supplement lack acknowledgment and that's only the tip of the iceberg.

Crop observing: Wasteful crop checking is an enormous impediment. With drones, time-series movements can show the advancement of a crop and uncover creation shortcomings, empowering better administration.

II. LITERATURE SURVEY

C. Zhou et al.,[1] presents a rebuilt lingering thick organization was proposed for tomato leaf illness recognizable proof; this mixture profound learning model joins the upsides of profound remaining organizations and thick organizations, which can lessen the quantity of preparing measure boundaries to further develop computation precision just as upgrade the progression of data and slopes. The first RDN model was first utilized in picture super goal, so we need to rebuild the organization design for grouping undertakings through changed information picture highlights and hyper boundaries. Exploratory outcomes show that this model can accomplish a best 1 normal recognizable proof

exactness of 95% on the Tomato test dataset in man-made intelligence Challenger 2018 datasets, which checks its agreeable exhibition

P. K. V et al.,[2] presents novel methodology and procedure used to advance the simple and straightforward execution of the program in recognition. The created model can distinguish thirteen totally various types of plant infections from sound leaves, with the adaptability to differentiate from the encompassing plant leaves. This strategy for the ID of infections was anticipated interestingly as indicated by our insight. All vital advances were taken by farming advisors to join this sickness acknowledgment model, starting with the assortment of photos to make subtleties. Python and PyCham are utilized to do the profound CNN measure.

D. Varshney et al.,[3] provides AI are regularly used to distinguish the influenced leaf pictures. The different AI calculations used to decide if a plant is contaminated or not with an illness, are examined in this investigation. It was done in different advances, for example, picture obtaining, highlight extraction, order of the disease and result show. This work additionally needs to do a precise investigation of different strategies for the distinguishing proof of illnesses of plants. The point is to recognize the plant sicknesses utilizing picture examination. It additionally, after discovery of the sickness, says the name of compost to be utilized. The vermin and creepy crawlies responsible for the pandemic are additionally depicted.

C. S, S. Ghana et al.,[4] presents inescapable dissemination of cell phones among crop producers all throughout the planet with a normal 5 billion cell phones by 2020 offers the capability of transforming the cell phone into an important device for different networks developing food. One potential application is the improvement of versatile sickness diagnostics through AI and publicly supporting. With the assistance of condition of workmanship advancements like profound learning and distributed computing, the equivalent can be accomplished on an ongoing premise.

P. Kanade et al.,[5] presents a layout of modified combinations of plant ailments and diverse social affair measures in computer based intelligence, which are used in different plant leaves for recognizing contaminations. The preliminary outcomes show that, even with a precision of 96.7 percent, the use of changed simulated intelligence procedures could be effectively used for the gathering of various plant leaf diseases. This method would be exceptionally helpful for the farmers to avoid crop hurt, the shortfall of food creation in the public field and the abuse of money on cultivating things.

H. D. Gadade et al.,[6] provides a precise and strong technique for tomato leaf illness distinguishing proof just as characterization into different phases of advancement utilizing AI is proposed. The work is completed in two phases. Right off the bat the tomato leaf pictures will be characterized into proper infection types. Then, at that point in the subsequent stage, the tomato leaf illness is analyzed at different phases of advancement. Recognizing the phase of advancement of tomato leaf would assist with choosing the sort and measure of treatment needed for the plant. The unhealthy leaf pictures which are taken from the PlantVillage dataset have been characterized into high, medium, low, and ordinary seriousness reviewing.

K. Kirti et al.,[7] proposed strategy yields better execution and exactness in recognizing the sickness, than past AI based methodologies. Grape Plant dataset from PlantVillage Information base is utilized for the work. The dataset contains all out 1807 pictures (sound and unhealthy). ResNet 50 engineering of Profound Neural Organization in mix with Move Learning and Adjusting was utilized to process the outcomes. The proposed framework gives a precision of over 97% and performed better compared to the current methodologies which depend on include extraction techniques.

Deepa et al.,[8] Plant infections are the normal justification low yield and diminished pay to the ranchers. Presently, specialists are making an honest effort to discover an instrument that consequently recognizes the plant infections. Exact ID of plant infections might help in discovering a cure at the most punctual to control the misfortune. This work endeavors to foster an original methodology by utilizing AI procedures to foresee the plant infections. Test results show that the plant sicknesses can be precisely arranged.

X. Li et al.,[9] utilized informational indexes of apple dark spot infection, dark star illness, cedar rust sickness and sound passes on to examine the distinguishing proof and characterization of apple leaf illnesses. Picture division SVM classifier and ResNet and VGG convolutional neural organization model were utilized for examination and

improvement. In the last analysis, ResNet-18 with less layers of ResNet got an exactness pace of 98.5% accomplishing better acknowledgment impacts.

S. Kumar et al.,[10] proposes a strategy for plant sickness recognizable proof through a bunch of assortment of leaf pictures. This strategy is created dependent on a profound convolutional autoencoder approach. Unique pictures are right off the bat reproduced by the autoencoder and afterward by utilizing just encoder part, the components of these pictures are extricated. The yield highlights are taken care of into Help Vector Machine for order. A little arrangement of Plant Town dataset is considered to check the exhibition of the recreated execution by the convolutional autoencoder. Then, at that point, for the arrangement stage, recreation results acquired have shown that the distinction of autoencoder design and parts of classifier give diverse normal precision. In this way, the most noteworthy precision came to 98.8% for plant illness arrangement.

H. Yu et al.,[12] proposed leaf spot consideration organization, the component division subnetwork is first educated with another picture set, where the foundation, leaf region, and spot region are explained. In this manner, the spot-mindful grouping sub network is associated with the component division subnetwork and afterward prepared through ahead of schedule and later combinations to create the semantic-level spot include data. The test results affirm that the proposed organization can build the discriminative force by displaying the leaf spot consideration component. The outcomes demonstrate that the proposed technique beats ordinary cutting edge profound learning models.

III. CHALLENGES

The challenges selected as the most impactful were the following:

- The background often contains elements that can make it very difficult to correctly segment the region of interest where the symptoms are manifest.
- Capture conditions are difficult to control, which may cause the images to present characteristics that are difficult to predict and make the disease identification more challenging.
- Most symptoms do not have well defined boundaries, rather gradually fading into normal tissue, making it difficult to clearly define which are the healthy and diseased regions.
- A given disease may possess very different characteristics depending on its stage of development, and sometimes on where it is located on the plant.
- Symptoms produced by different diseases may be present simultaneously, manifesting either physically separated or combined into a “hybrid” symptom that may be difficult to identify.
- Symptoms produced by different diseases may be visually similar, which forces the methods to rely on very tenuous differences to discriminate among them.

There are some other challenges that affect automatic disease identification that cannot be categorized together with those already discussed.

- The first is real-time operation. Although all proposed methods are expected to operate under certain time constraints, only a few applications actually require realtime operation.
- Since the computational power available continues to grow, meeting real-time requirements would be expected to be easier with time.
- However, the resolution of the images also increases, thus demanding more computational resources. Also, portable devices and low cost computers such as Raspberry Pi (Raspberry Pi Foundation, Cambridge shire, UK), have limited resources. Thus, depending on the intended application, reducing computational complexity and memory requirements may be a major concern. Using efficient programming languages, structuring the code to avoid unnecessary stress on computational resources, and reducing the image resolution during the processing are among the most common solutions adopted for reducing computational requirements.

IV. PROPOSED METHODOLOGY

The work starts with to define input configuration and taken plant disease dataset. The methodology steps is as followings,

1. **Input:** The dataset Leaf image are implemented as input. The input image are taken in the format .jpg or .png
2. **Preprocessing:** The collected input images are subjected to preprocessing. In the Preprocessing step we can implement image rescaling.
3. **Feature Extraction:** In the feature extraction process, the plant disease pattern features are extracted.
4. **Feature Optimization:** In this process, the features are optimized by means of the Spider Monkey Optimization (SMO) algorithm.
5. **Classification:** In this step, the Support Vector Machine (SVM) is used to identify the type of leaf disease.
6. **Performance Estimation:** In this module, the performance metrics like Accuracy, sensitivity and error rate is estimated.

V. CONCLUSION

The use of digital image processing and computer vision in plant diagnosis through machine learning is still new, which means there are still many alternatives to be explored with the potential to minimize at least some of the issues pointed out herein. Additionally, with the availability of greater computational power, strategies that were previously prohibitive may now be applied. Finally, advancements in imaging imply that images with superior quality can now be captured at low costs, and new improvements will certainly be developed. Future work will be focused on to implement machine learning based adaptive algorithm to identified plant leaf diseases.

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