Deep Learning for Medicinal Plant Identification and Utilization: Leveraging ResNet for Enhanced Recognition and Applications

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Abstract

In addition to their many medicinal uses, herbal plants also have the added benefit of releasing oxygen into the air at no cost to the environment. There are a lot of medicinal plants that are also helpful for future generations since they contain active ingredients. A lack of understanding of medicinal plants, together with issues like climate change, population increase, occupational secrecy, and inadequate government funding for study, are causing the extinction of many important plant species. Current algorithms sometimes struggle to accurately identify herbal leaves throughout the year because of the latency of dimensions parameters like length and breadth. Therefore, to enhance the detection rate for herbal leaf identification, the suggested approach zeroes in on the incomplete dataset issues. In this research, we provide a deep learning method that makes use of a CNN trained on the VGG-16 and VGG-19 models. The picture segmentation procedure is yielding excellent results with the addition of dimension parameters in the datasets. The process of using a machine learning classifier in conjunction with ex-orgate operations to confirm the acquired result is known as deep knowledge-based identification. The detection rate of herbal leaves is being improved by this two-stage authentication (TSA) technique. For picture segmentation, we used ResNet as well. As a result of combining image segmentation with machine learning, the proposed architecture is becoming more robust. Also, the detection accuracy is being improved by using intelligent selection of images segmentation techniques to segment the leaf from the image. Deep Learning, VGG 19, Herbal Plant Identification

I. INTRODUCTION

Since ancient times, indigenous communities have relied on herbs as traditional medicines.Plants play a vital role in our planet for many reasons. Herbalists often use their years of sensory experience to choose which plants to use. Technological advancements in analysis have greatly aided bald face identification.Particularly for individuals who have never dealt with herbal recognition before, this is a huge help. Testing in a controlled environment necessitates expertise in sample handling and data interpretation, on top of the time-consuming processes. It is now more important than ever to be able to reliably and simply identify the leaves of medicinal plants. It is quite probable that a robust method for precisely categorizing plants is computing using probability and statistical analysis. The identification of herbal leaf choices is a suggested approach for rapidly and nondestructively identifying herbal leaf items. People who are unable to purchase costly analytical tools might benefit greatly from this strategy. We are all familiar with the following herbal leaves: Tulsi, Oma Valli, Neem, Vana Thulai, Thudhuvalai Lime, and so on. Sometimes, the shape and color of the herbal plants have been employed as features in the categorization process.Plants are categorized using several classification techniques that are based on functional vectors. In order to build a very precise approach for herbal plant identification, the Support Vector Machine (SVM) algorithm has been the subject of many recent research articles. The distribution of Gauzian leaf characteristics will give greater efficient classification, according to recent discussions on algorithms. Additionally, the study papers will go deeper into the research methodologies used to achieve improved classification via the implementation of several algorithms, including probabilistic statistical neural networks, supporting vector machines, main component analysis, and texture base analysis, all of which are utilized for plant detection. Because they are constantly accessible and have just two dimensions, many writers rely on plant leaves alone when attempting plant identification. Knowing the cultural importance, nutritional value, and therapeutic qualities of herbal plants is essential for identifying and making use of them. Natural cures for a wide range of medical issues may be found in these plants, which also happen to be excellent nutritional sources. They are also employed in rites and rituals throughout many different cultures, which adds to their cultural importance. The protection of these plants and their environments is guaranteed by sustainable harvesting procedures, which in turn promote biodiversity and environmental sustainability. By incorporating herbal plants into our lives, we may access a more complete view of health and wellness, paying homage to traditional knowledge while encouraging a more inward orientation.

II. LITERATURE SURVEY

Pharmaceutical Due to their lower cost and lack of side effects compared to existing medications, medicinal plants are being used more and more by enterprises. Consequently, many researchers are keen to learn more about automated medicinal plant categorization. It is necessary to develop a robust classifier capable of real-time accurate categorization of medicinal plants. Many deep learning and machine learning methods have been used to classify plants from images of their leaves in recent years; this article examines their efficacy and predictability.[1]. Natural antioxidants are now getting a lot of attention, and there are attempts to substitute synthetic ones. Natural antioxidants like these may also aid in the prevention of oxidative damage by being added to functional meals. Developed nations aren't the only ones considering a shift in medicine consumption patterns as a result of chemical plant adverse effects. [2]. This essay describes an advocacy project as a community psychology enactment and traces the status of TRM from colonial to modern times. Community psychology's focus on social justice, empowerment, cooperation, empirical foundation, and community mobilization is in line with the goals of the enactment, which is to formally include TRM into basic healthcare services in Ghana.Specifically for medicinal plants unique to Borneo, the study [4] suggests a strategy for automated, real-time plant species detection. In order to overcome the recognition difficulty, a refined EfcientNet-B1 model was trained and evaluated using both publicly available and privately owned plant species datasets. The suggested model outperformed the standard model on test data by more than 10% in both datasets. Research on the viability of automatic recognition for the accurate identification of medicinal plants was the driving force behind the work [7]. The authors of this research provide a new database of medical herbs that contains pictures of 11 different types of plants, 10 of which are medicinal and 1 of which are not. Next, we used the MobileNetV3 architecture to suggest a model for the efficient, accurate, and cost-effective categorization of medicinal plants.Differentiating between several kinds of Indian leaves using methods based on convolutional neural networks (CNNs) is examined in the research [8]. A method for identifying plant species is given in [9]. Additional information on vegetation may be extracted from photos with the use of ExG-ExR, an improved vegetation index. Setting a threshold value for OTSU is superfluous since it already has a zero threshold. Their ExG-ExR index works well regardless of the lighting background, even if ExG using Otsu's method generates more fora and fauna information. in [10], A deep convolutional neural network (CNN) architecture might be built and its settings tweaked to improve the accuracy of recognition. Findings from this study stress the need of considering the effects of the multi-layer technique on small samples when designing effective algorithms. Data augmentation also has more significant beneficial effects on output. The research [11] looks at how six different medicinal species are classified based on various morphological, color, and textural traits. Near the Indian state of Assam, they photographed 90 leaves from 6 distinct medicinal plants. Prior to its combination for increased accuracy, each feature was used alone for the categorization task. An effective method for identifying leaves is the back propagation neural network, or BPNN. The suggested strategy is shown to enhance accuracy in extensive testing on 90 photos of six different kinds of leaves. Quick species identification from leaf photos is made possible using an ensemble learning method to the DL topic in artificial intelligence, as shown in research [12].[15]Image resizing is an essential part of computer vision pre-processing. When training on smaller pictures, deep learning models mostly benefit. The training time for the architecture increases as the input picture size increases since the neural network has to learn from four times the number of pixels. Here, we detailed how the impact of picture scaling on training time and performance for models has changed over time. don't forget.

i) Proposed Work:

III. .METHODOLOGY

In addition to their many medicinal uses, herbal plants also have the added benefit of releasing oxygen into the air at no cost to the environment. There are a lot of medicinal plants that are also helpful for future generations since they contain active ingredients. A lack of understanding of medicinal plants, together with issues like climate change, population increase, occupational secrecy, and inadequate government funding for study, are causing the extinction of many important plant species. Current algorithms sometimes struggle to accurately identify herbal leaves throughout the year because of the latency of dimensions parameters like length and breadth. Therefore, to enhance the detection rate for herbal leaf identification, the suggested approach zeroes in on the incomplete dataset issues. A convolutional neural network (CNN) trained on the VGG-16 model is suggested as a deep learning-based method in this research.

The picture segmentation procedure is yielding excellent results with the addition of dimension parameters in the datasets. The process of using a machine learning classifier in conjunction with ex-orgate operations to confirm the acquired result is known as deep knowledge-based identification. The detection rate of herbal leaves is being improved by this two-stage authentication (TSA) technique. For picture segmentation, we used ResNet as well. As a result of combining image segmentation with machine learning, the proposed architecture is becoming more robust. Also, the detection accuracy is being improved by using intelligent selection of images segmentation techniques to segment the leaf from the image.

ii) Architecture of the System:

Several essential parts make up the system architecture that attempts to forecast medicinal plants. The dataset is first examined and prepared using data exploration and preprocessing methods. In order to build and evaluate models, the dataset is split into two parts: the training set and the testing set. A number of ML and DL models do a great job at learning intricate plant picture representations, which allows for precise identification and categorization. Models trained on massive plant picture datasets are able to differentiate between species in a wide range of environments by examining minute visual clues. With this skill, plant identification procedures may be made more efficient and easily integrated into fields like pharmacology, agriculture, and biodiversity protection, among others. To further facilitate their long-term use in herbal medicine and medication development, deep learning methods provide light on the biochemical make-up and therapeutic effects of medicinal plants. Deep learning has the ability to transform our knowledge of medicinal plants for the benefit of humans and the environment by combining cutting-edge computational methods with botanical expertise.

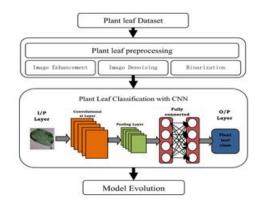


Fig 1: Proposed Architecture

iii) Dataset:

Various species of medicinal plants are included in the dataset used for this investigation. The pictures came from a variety of places, such online resources, herbaria, and botanical gardens. Eighty percent of the photos were used for training, while the remaining twenty percent were used for testing.



iv) Algorithms:

CNN:

Deep learning network architectures that learn from data directly include convolutional neural networks (CNNs or ConvNets). In order to identify objects, classifications, and categories, CNNs excel in detecting patterns in pictures. When it comes to audio, time-series, and signal data classification, they can also be very successful.

One strong model for image classification tasks including object identification, face recognition, and medical picture analysis is ResNet50, a residual network that has been trained on the massive ImageNet dataset. ResNet50 achieves an error rate that is on par with human performance.

Another well-known deep convolutional neural network (CNN) for picture identification is the VGG-16 model. When it comes to computer vision tasks like object identification, segmentation, and picture classification, VGG-16 is a popular choice for feature extraction or as a pre-trained model for transfer learning. Its ease of use and high performance have made it a favorite among deep learning researchers and practitioners. The Visual Geometry Group (VGG) of Oxford University created the VGG-16 design, then they built an expansion of it called VGG-19. With 19 layers, including 16 convolutional and 3 fully linked ones, it goes deeper than VGG-16.

IV. EXPERIMENTAL RESULTS

Accuracy: Find out how well the deep learning model does in general at identifying medicinal plants in pictures. This is the fraction of the total number of photos in the dataset that were properly categorized. Bring forth the deep learning model's total recall for all plant species. This is the mean percentage of class-wide forecasts that turned out to be correct.

Accuracy: Offer the total accuracy of the deep learning model for all plant species. From all cases projected as positive, this is the average fraction of forecasts that turned out to be correct.

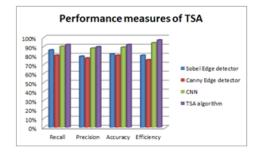


Fig 3: Data Analysis

Performance Evaluation

S.No	Methods	Recall	Precision	Accuracy	Efficiency	Recognition error (%)	
1	Image Segmentation process (Sobel Edge operator)	86%	79%	81.5%	80%	0.071	
2	Canny Edge detector	80%	77%	80%	75%	0.175	
3	Single classifier (CNN)	90%	88%	89%	94%	0.053	
4	Proposed TSA algorithm	92%	89.5%	92%	97%	0.008	

Fig 4: Performance Evaluation



Fig 5: Home Page



Fig 6: Upload File

						Form
						Choose File No file chosen
						Upload
1	:	:	;	:	:	

Fig 7: Choose File



Fig 8: Predicted Result



Fig 9: Predicted Result

V. CONCLUSION

Therefore, the suggested approach has successfully detected the leaf of the herbal plant. Through twostage verification methods, it has shown that the detection rate is superior to other pure image processing approaches and single toneclassifiers. Our texture and multispectral datasets will aid in feature selection for medicinal plant optimization. The medical laboratory has standardised and polished this dataset to a high degree. As part of our method, we use our standard dataset to pick features using chi-square analysis. Also, the two-stage verification procedures that we suggest using to identify the herbal plant leaves are our own creation. By using our suggested methods, we were able to increase the detection rate in the categorization stage. Farmers and regular people may use this to their advantage and boost Ayurveda provision production. Businesses in the fields of botany, taxonomy, Ayurveda manufacture, and Ayurveda practice may benefit from this flawless automated categorization system, which aids in the identification of medicinal plants without the need for human assistance. Additional medicinal plant parts, including flowers, fruit, and seeds, will be accurately identified and classified using our suggested algorithm. When compared to previous technologies, this edge-based detection of medicinal plants is more accurate. Even while our suggested solution is more efficient, it requires a lot of processing power and is pixel-based.

VI. OUTLINE FOR THE REST OF THE WORK

We focused on supervised learning in this study, using Deep Learning with CNN and ResNet 50 algorithms to identify and use herbal plants; to improve our model's accuracy, we will implement an unsupervised learning technique on top of deep learning with different algorithms in the future.

REFERENCES

- [1]. Chanyal H, Yadav RK, Saini DKJ. Classification of medicinal plants leaves using deep learning technique: a review. Int J Intell
- [2]. Syst Appl Eng. 2022;10(4):78–87.
- [3]. Javid A, Haghirosadat BF. A review of medicinal plants effective in the treatment or apoptosis of cancer cells. Cancer Press J. 2017;3(1):22–6.
- Barimah KB, Akotia CS. The promotion of traditional medicine as enactment of community psychology in Ghana. J Community Psychol. 2015;43(1):99–106.
- [5]. Rao RU, Lahari MS, Sri KP, Srujana KY, Yaswanth D. Identi fcation of medicinal plants using deep learning. Int J Res Appl Sci Eng Technol. 2022;10:306–22.
- [6]. Singh V, Misra AK. Detection of plant leaf diseases using image segmentation and soft computing techniques. Inf Process Agric.
- [7]. 2017;4(1):41–9.
- [8]. Malik OA, Ismail N, Hussein BR, Yahya U. Automated realtime identification of medicinal plants species in natural environment using deep learning models—a case study from Borneo Region. Plants. 2022;11(15):1952.
- [9]. Valdez DB, Aliac CJG, Feliscuzo LS. Medicinal plant classification using convolutional neural network and transfer learning. In: 2022 IEEE International Conference on Artificial Intelligence in Engineering and Technology (IICAIET). IEEE; 2022. p. 1–6.
- [10]. Abdollahi J. Identification of medicinal plants in ardabil using deep learning: identification of medicinal plants using deep learning. In: 2022 27th International Computer Conference, Computer Society of Iran (CSICC). IEEE; 2022. p. 1–6.
- [11]. Sivaranjani C, Kalinathan L, Amutha R, Kathavarayan RS, Kumar KJJ. Real-time identification of medicinal plants using machine learning techniques. In: 2019 International Conference on Computational Intelligence in Data Science (ICCIDS). IEEE; 2019. p. 1– 4.
- [12]. Zin IAMd, Ibrahim Z, Isa D, Aliman S, Sabri N, Mangshor NNA. Herbal plant recognition using deep convolutional neural network. Bull Electr Eng Inform. 2020;9(5):2198–205.
- [13]. Saikia AP, Hmangaihzuala PVL, Datta S, Gope S, Deb S, Singh KR. Medicinal plant species classification using neural network classifer. In: 2021 6th International Conference on Communication and Electronics Systems (ICCES). IEEE; 2021. p. 1805–11.
- [14]. Sachar S, Kumar A. Deep ensemble learning for automatic medicinal leaf identification. Int J Inf Technol. 2022;14(6):3089-97.
- [15]. Manoharan JS. Flawless detection of herbal plant leaf by machine learning classifer through two stage authentication procedure. J Artif Intell Capsule Netw. 2021;3(2):125–39.
- [16]. Saponara S, Elhanashi A. Impact of image resizing on deep learning detectors for training time and model performance. In:

Applications in Electronics Pervading Industry, Environment and Society: APPLEPIES 2021. Cham: Springer International Publishing; 2022. p. 10–17.

- [17]. Shorten C, Khoshgoftaar TM. A survey on image data augmentation for deep learning. J Big Data. 2019;6(1):1-48.
- [18]. Sae-Lim W, Wettayaprasit W, Aiyarak P. Convolutional neural networks using MobileNet for skin lesion classification. In: 2019 16th International Joint Conference on Computer Science and Software Engineering (JCSSE). IEEE; 2019. p. 242–7.
- [19]. Wang W, Hu Y, Zou T, Liu H, Wang J, Wang X. A new image classification approach via improved MobileNet models with local receptive feld expansion in shallow layers. Comput Intell Neurosci. 2020.