

Identification of Ayurvedic Medicinal Plant Using Deep Learning

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Abstract: Ayurvedic medicine is ancient medicine. This therapeutic approach makes use of plant materials that are used in Ayurvedic medicine. The plants need to be identified because they differ from the many other plant species that can be found in nature. Without the proper knowledge, it could be difficult for the typical person to identify locally available herbal remedies. This demonstration shows a new technique that uses convolutional neural networks (CNN) and leaf images to identify the leaves of Ayurvedic medicinal plants. Computer technology advancements have allowed the field of computer vision to expand to include a wide range of applications. One of its applications is image classification, where it recognizes images more accurately than traditional methods. This document contains all of the information and direction needed to complete each step of the implementation process. All of the basic steps are covered in great detail, including building a database by gathering images and training models. Compared to other methods, our deep neural network method yields a more accurate classification. Another benefit is easier feature extraction from the image, which can be fed into the model without requiring preprocessing. One way to feed deep convolutional neural networks is with raw photo data. Without needing to extract the leaves themselves, we can precisely classify leaves using deep neural networks, which capture and store visual properties as an image moves through several layers. Web applications and deep learning are used to sort and present worksheets. The deep learning technology used in this essay is the convolutional neural network.

Keywords: image pre-processing, rectified linear units, deep learning, convolutional neural networks, and ayurvedic medicinal plants.

Introduction:

Ancient medicine is Ayurvedic medicine. Nature

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has surrounded us with lot of plants having medicinal values in the modern day, its significance has returned. Plant materials employed in Ayurvedic medicine are used in this therapeutic approach. Because they are different from the numerous other plant species that exist in nature, these plants must be identified. It might be challenging for the average person to recognize locally accessible herbal remedies without the right information. This demonstration presents a novel method for detecting Ayurvedic medicinal plant leaves using convolutional neural networks (CNN) and leaf pictures.[1]

Plants have been used therapeutically in many different fields. To set them apart is a challenging task. You will find that it is easier to use this technology. A survey of the literature found that up to 80% of Indians follow traditional medicine, which includes Ayurveda. Many sources state that 75–80% of patients use herbal treatment as their main kind of care.[13]

Ayurveda, a traditional medicine practiced by up to 80% of Indians, is just one field of plant-based medicinal use. Ayurveda is considered to be the oldest healing science. In Sanskrit language,

Ayurveda means “The Science of Life”. The complex task of differentiating between the various therapeutic plants can be alleviated with specialized technology. Herbal medicine is the primary form of treatment for an estimated 75-80% of people. Neurons have multiple layers and the inputs for the next layer come from connected nodes and their outputs. It is a difficult task to classify medicinal plants, but with the help of technology, it can be made more manageable. With three layers, the neural network comprises the input, hidden, and output layers. Extracting data from the input involves running simulations through all layers of the neuron. A deep neural network is one that has several layers and neurons. While standard neural networks have their limitations, deep learning shows itself to be more adept at solving complex problems. Deep learning has eliminated the need for the previously necessary extraction of visual features. Deep learning uses an RGB value matrix to perform specific computations on an image in order to extract features.[3]

The world is home to thousands of different plant species, some of which are toxic to humans, some of which are endangered, and some of which are used in medicine. The origins of Ayurveda date back over 5,000 years. Not only are plants a vital source of food for humans, but they also form the foundation of all food chains. Plants must be appropriately inspected and categorized in order to be used and protected. Accurate identification of unknown plants is dependent on the knowledge and expertise of professional botanists. The best manual methods for straightforward and accurate plant identification are those that use morphological traits. Thus, in many of the involved processes, the classification of these plant species "depends on the accumulation of human knowledge and skill". Unfortunately, this manual detection process is time-consuming and labor-intensive. Consequently, a great deal of research has been done by scientists to help with the automatic classification of plants based on their physical characteristics.[4]

As an essential component of our ecology, plants are becoming fewer in number, which is cause for great concern. Although the processes are very similar, the systems that have been developed so far automate the classification process using varying numbers of steps. These procedures basically consist of getting ready for the leaves to be collected, preprocessing them to determine their unique characteristics,

classifying the leaves, adding data to the database, training the database for recognition, and assessing the outcomes. Although stems, flowers, petals, seeds, and even entire plants can be utilized in automated processes, leaves are the most widely used feature for plant identification. Automated plant identification systems make it quick and simple for non-plant experts to identify different species of plants.[5]

Plants generate O₂, which is essential to all life on Earth. Plants are essential to preserving the diversity of life on Earth because they provide both water and oxygen, despite their diverse range of forms and sizes. Ayurvedic medicines and medicinal plants are used to cure a wide range of human ailments. Ayurveda is used in many different forms to heal human illnesses. These plants have extraordinary restorative properties right down to their roots. People employ plants in their everyday lives to make food, medications, and cosmetics

[14] There are many types of medicinal plants, some of which are difficult to classify because of their similarities. Therefore, Ayurveda users must distinguish them. In some countries, most specialists still use traditional methods to manually distribute herbs. Most of the people in the world rely on traditional medicine which is made from medicinal plants. Medicinal plants have characteristics useful for human and animal health. In past years medicine were initially called simple plants, but today they are known as herbal plants.[6]

The plant is hardly used as a whole; minimum one of its components is used to make botanical medicine. Distinct parts of the same plant can be used for different purposes. Plants with healing properties can also be used in feed or even made into clean drinks. People have always looked towards nature for ways to heal diseases between 65% and 80% of people on the planet use medicinal plants as remedies for a range of illnesses. Preceding to the development of medical chemistry in the 16th century, to treat and prevent many diseases medicinal plants has been used.[7] However, because of the reducing effectuality of synthetic drugs and the extending number of side effects associated with their use, the use of natural medicinal plants has returned to the spotlight. Plants are the most essential natural habitat for various organisms. In Addition, innumerable of those whose fuels such as coal and conventional natural gas today use plants that have been around for a long time.

However, humans have worked on plant ecosystems in recent years, that means many crops can no longer be grown on the other hand, the ensuing environmental disaster have produced a series of damaging effects such as, weather anomalies, and earth tremor, threatening human life and development. The primary ingredients of ayurvedic medicines are plant leaves, along with other plant parts like bark, roots, and so forth. The herbal manufacturing is rife with substandard material that danger human well-being and threaten global growth.[8]

Therefore, develop medicinal classification method has been a heated region of research. It is now generally accepted that the leaves of the plants have characteristics that makes them simple to separate and evaluate. Therefore, it is chemical-free and it is used as the primary means to recognize all herbal plants. The application of automatic computer image identification in this industry is expanding because of how quickly images can be processed. AyurLeaf, a Convolutional Neural Network model based on Deep Learning, is proposed in this work to categorize medicinal plants.[9]

Literature Survey:

The authors of paper [1] developed a system that will help with plant identification and provide information about its medicinal properties in an effort to encourage the use of natural remedies for a range of illnesses. The Support Vector Machine algorithm is utilized to classify the datasets after they have been gathered and their features extracted using texture and HOG.

The authors of paper [2] used machine learning and computer vision to classify medicinal plants found in the Western Ghats. The properties produced by the KNN, HOG, and SURF classification algorithms are utilized in the procedure. HOG features are used to design leaf veins, while the SURF feature descriptor models 20 different points of interest for leaves. This technique classifies leaves that are k times l using the K-NN algorithm. The extraction process of this model is computationally expensive, even though K-NN classification is more accurate than 96% of the time.

Botanists can now make use of a model developed by the authors of the paper [3] for digitizing plant specimens. They provided assistance in maintaining large-scale photos of Plummer flora plant species. An electronic field guide system prototype was

developed, enabling users to use their mobile devices to browse a sizable photo library. They offered effective algorithms for compiling information on different plant species.

The goal of the paper's authors [4] was to categorize and verify the herbs and other medicinal plant components that are frequently used in Indian herbal remedies. Only after these leaves' consistency and authenticity have been verified can herbal remedies be made. The medicinal plant's leaves were meticulously sorted and examined. Lastly, a comparison was made between the leaf's precise texture and group measurements and the database. This strategy is employed due to the way it looks.

Using images of plant leaves, the authors of paper [5] created a classification system based on image processing. The program displays the query's closest match. The competency of the system is assessed using ten distinct plant species following the implementation of the suggested algorithm. The software is tested with 50 leaves (differing plant species) after being trained with 100 leaves (10 of each species). 92% of the suggested algorithms are found to be successfully used.

According to paper [6], by combining multiple color image channels, the Multichannel Modified Local Gradient Pattern (MCMLGP) feature descriptor was created as a novel way to improve classification performance with more significant feature extraction. The authors' proposed method was trained on an SVM classifier with HI, linear, and polynomial kernels. On our dataset of medicinal plants, various feature descriptors were tested for comparison experimental analysis using MCMLGP. With an accuracy rate of 96.11%, the results demonstrated the superiority of the suggested approach over alternative techniques. Because of its effectiveness, it is a useful tool for researching and developing the classification of medicinal plants.

By removing form, color, and texture characteristics from leaf pictures, the authors of study [7] proposed a technique for correctly classifying leaves using an artificial neural network classifier. Choosing best picture input feature is critical for maximizing efficiency while minimizing computational complexity. They tested the network's accuracy using various combinations of input features. According to test findings on 63 leaf photos, this approach can reach 94.4% accuracy when at least 8 input features are used. This technique is more commonly used in

leaf identification systems with low input and calculation requirements.

The authors of the research [8] created a model (Deep Neural Networks) for identifying therapeutic plants. In publication [7], The scientists outlined a process for extracting form, color, and texture information from leaf pictures so that the type of leaf could be properly identified through training an artificial neural network classifier.

The authors of study [9] developed AyurLeaf, a Deep Learning-based Convolutional Neural Network (CNN) model, to categorize medicinal plants according to leaf characteristics like size, shape, color, texture, and so forth. This study also proposes a standard dataset for medicinal plants that are commonly found in Kerala, a southwest-coast state of India. For forty medicinal plants, leaf samples are included in the proposed dataset. A deep neural network based on the Alex net successfully retrieves the features of the dataset. To complete the classification process, SVM and Softmax classifiers are employed. Our 5-cross validation model produced a 96.76% classification accuracy using the AyurLeaf dataset. With AyurLeaf, you can easily categorize and identify medicinal plants, which helps to preserve traditional medical knowledge. In paper [10], the authors offered a fresh and efficient method for getting leaves. The VGG-16 feature map is then created by converting the image to a device-independent color space. The goal of re-projecting this feature map to the PCA subspace is to improve the performance of species recognition. The work uses two different datasets of plant leaves to demonstrate the robustness.

Methodology:

A. Method of data collection :

Without a doubt, the dataset is the most important component in projects involving deep neural networks or object recognition. Many details about the data in question can be found in a well-curated dataset. Images used in this particular project came from datasets collected on Kaggle. A Python script that made use of the comparison technique was employed to eliminate any duplicate images present in the dataset in order to eliminate repetition. The text in the pictures was enhanced with contrast while taking into account the names, sizes, and dates of the images.

Sorting images into distinct classes according to

their categories comes after duplicate have been completely removed. There are datasets made for testing, validation, and training. All the varieties of medicinal leaves are present in every dataset. The model is routinely trained using the training dataset and tested using the test dataset to guarantee that the output is accurate, and validated using the validation dataset. Thirty distinct kinds of leaves are collected: Jamun, Tulsi, Neem, Rasna, Jackfruit, Basale, Indian Mustard, Karanda, Lemon, Peepal, Jasmine, Mango, Mint, Drumstick, Curry, Parijata, Betel, Mexican Mint, Indian Beech, Guava, Sandalwood, Rose apple, Fenugreek and many other plant leaves. In our dataset there are total 1835 Images. 80% of images are given for training and remaining 20% is again divided into two parts as 10% for validation and 10% for testing.

B. Method of Data Analysis :

1. Image Processing :

When we discuss the most basic form of image manipulation in which input and output are represented as frequency pictures. We are talking about image processing. One method to enhance image quality is to pre-process the image, which involves removing unwanted distortions or enhancing specific visual elements that may need further processing. Pre-processing methods include low-frequency ambient noise reduction, pixel-by-pixel equalization, reflection removal, picture intensity adjustments, and masking specific areas of the image. These methods frequently take advantage of surrounding pixels to enhance images and facilitate the perception of information by people. They also provide improved input for automated image processing systems.

Noise reduction is an important step in the image processing process because all high-frequency components in an image are considered noise. Filtering, denoising, and noise reduction processes all make use of low pass filters. Image filtering is one way to lessen the effects of missing, incorrect, and camera noise-related pixel values. After preprocessing, the image is now prepared to be fed into the algorithm.

2. Neural Network training :

Deep learning is easier to understand when one has the context that ml offers. Self-learning and self-developing computer algorithms are investigated in this field. Deep learning makes use of CNNs, which

are made to resemble human analysis and learning processes, whereas machine learning makes use of simpler ideas. The maximum throughput of earlier neural networks limited their complexity. Computers are now faster than humans at monitoring, comprehending, and reacting to complex events as a result of advancements in large data analysis. Denser and more intricate neural networks are created in order to do this. Desktop vision, image categorization, Deep learning has been beneficial for and speech recognition. Without assistance from humans, it can resolve any pattern familiar issue. Feature extraction is an additional facet of deep learning that use algorithms to automatically produce pertinent Multiple layer perceptron's and regularized CNNs are interchangeable. In general, multilayer perceptron's are fully linked neural networks that have connections between the neurons in the 1st and 2nd layers as well as between each layer's neurons. These neurons are susceptible to data overfitting because of their "finish connectivity". One popular regularization method is to include size verification of the weights in the loss function. CNNs use the data's hierarchical structure to construct larger, more complicated patterns from smaller, simpler ones in an effort to accomplish various forms of regularization. CNN is therefore ranked lowest in terms of complexity and connectedness. CNN's structural components are max pooling, fully connected layers, and convolutional layers. Typical design elements include a stack of several convolutional layers, one or more fully connected layers, and pooling layers. An essential part of the CNN architecture, the convolutional layer extracts feature by fusing linear and nonlinear techniques like the activation and convolution functions. Convolutional layers restrict the overall number of parameters by using the parameter sharing approach.

The convolutional layer, a crucial component of the CNN architecture, combines linear and nonlinear methods, such as activation functions and convolution, to extract features. Convolutional layers restrict the total number of parameters by using the parameter sharing strategy.

The non-linear layer of neurons is responsible for a variety of activation tasks. In multilayer networks, nonlinearities are desirable, and these functions introduce them. Other well-liked activation functions include RELU, tanh, and sigmoid. given that neural Linear Units (RELU) are preferred over

other features. Saturation nonlinearities are modified using Rectified Linear Units (Re LU). This activation function improves accuracy over time by gradually learning the rectifiers parameters at a low computational cost.

A pooling layer decreases in-plane dimensionality, improves linearity to min shifts and distortion, and minimizes the number of learnable parameters by applying a standard residual blocks operation to the feature maps. While all of the pooling layers have learnable parameters, filter size, stride, and padding are hyper-parameters in pooling operations, much like in convolution processes. Both local and global pooling layers are used by convolutional networks to combine the output of neuron clusters. You can use overlapping pooling to loosen up.

The neural network implements advanced reasoning through fully connected layers, which come later multiple convolutional and max pooling layers. Every activation from the stage before is connected to a neuron in a fully linked layers of a conventional neural network. Thus, an affine transformation plus a bias offset can be used to estimate their activations.

Through fully connected layers—which come after multiple convolutional and max-pooling layers, the neural network implements advanced reasoning. A neuron in a fully linked layer of a conventional (non-convolutional) artificial neural network is connected to each activation from the preceding stage. Thus, an affine transformation plus a bias offset can be used to estimate their activations.

Following the application of a filter, an image is obtained for the model. The image is then subjected to convolution, which is the result of multiplying each pixel value by the filter value, by the Convolution layer. The filter's size is another name for its depth. The size of the convolution layer's output is determined by the filter's depth. Stride-valued filter convolution is applied to the image. Each layer's neurons store the features of the image, and padding maintains edge information. The pooling layer, which controls pooling, receives the image after convolution. Layer images proceed to the fully linked layer after pooling, where the picture matrix is converted into a 1D array. CNN analysis does not necessitate lowering image resolution because multiple layers and filters already handle it. The pooled feature map needs to be flattened after it is obtained. The neural network receives a single column containing the entire pooled feature map

matrix, which it then processes.

Layer images move on to the fully linked layer after pooling, where the image matrix is converted into a 1D array. There's already a number of layers and filters taking care of it, so there's no need to lower the image resolution for CNN analysis. Once the pooled feature map is obtained, it must be smoothed. Once the pooled feature map matrix has been reduced to a single column, the neural network receives the entire matrix for processing.

After being pooled, the layer images are moved to the fully linked layer, where the image matrix is converted into a one-dimensional array. Multiple layers and filters take care of the image's resolution, so we don't need to reduce it for CNN analysis. After it is obtained, the pooled feature map needs to be smoothed. The neural network receives the pooled feature map matrix after it has been reduced to a single column for processing.

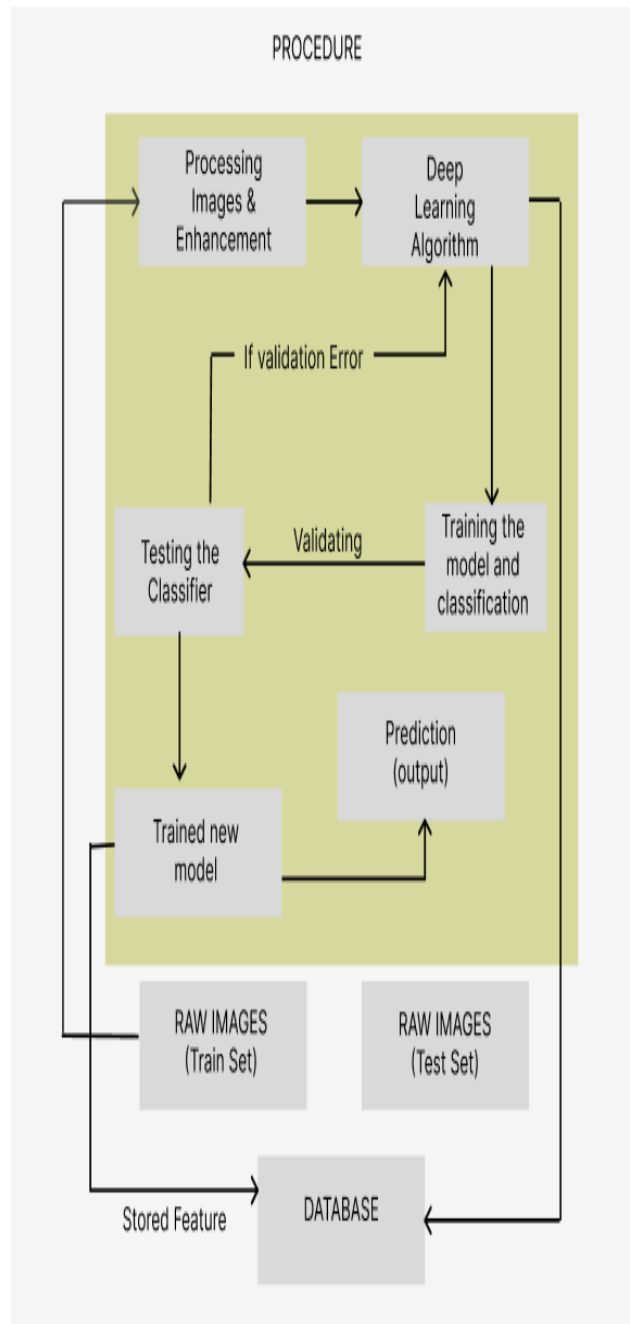


Fig.1. Architecture Diagram

Stride controls how the filter is distributed throughout the image's spatial dimensions. As a result, there are numerous output and overlapping receptive fields among the columns. Conversely, higher strides result in less overlap between the receptive fields, reducing the output volume as well as the spatial dimensions. Zero padding is the process of appending zeros to an image's dimensions in order to protect potentially lost data.

Proposed System :

There has been a proposal for a novel approach to medicinal plant identification. It uses photos of the leaves front and back portions taken from various perspectives. The database that served as the basis for this work included pictures of medicinal plant

I. Steps

1. Import Libraries

Import the TensorFlow library.

Import necessary modules including layers, models, and pyplot

2. Set Constants

Set the batch size as 32. Set the image size as 256.

Set the number of channels as 3 (assuming RGB images). Set the number of epochs for training as 50.

3. Load Dataset

Specify the path to the dataset.

Use the `image_dataset_from_directory` function to load the dataset.

Shuffle and batch the data.

4. Visualize Dataset

Display a sample of the loaded dataset for visual inspection.

5. Split Dataset Make training, validation, and test sets out of the dataset.

10% should be used for validation, 10% for testing, and 80% for training.

6. Preprocess Dataset

Cache, shuffle, and prefetch the training, validation, and test datasets for improved performance during training.

7. Build Model Create a Convolutional Neural Network (CNN) model.

Define layers for resizing and normalization.

8. Data Augmentation

Implement data augmentation techniques to enhance model generalization and performance.

9. Apply Data Augmentation

Apply the data augmentation techniques to the training dataset

10. Compile Model

Utilize the Adam optimizer, the Sparse Categorical Cross entropy loss function, and accuracy as a metric to compile the model.

leaves. Combinations of distinctive morphological characteristics, like texture and shape, have been found to increase the identification rate of green leaves. The system can identify whether a given image of a leaf from any plant is from a medicinal plant or not using this method. It shows a picture of the leaf with details about its properties, common and scientific names, and potential uses in medicine. Because of its many appealing advantages, this method uses the Dense Net kind of Convolutional Neural Network (CNN), which include promoting feature reuse and strengthening feature propagation, which both improve efficiency and reduce valuation loss. In this case, the model is trained using Keras and data.

11. Train Model

Train the model on the training dataset, validating on the validation set at the end of each epoch.

12. Evaluate Model Evaluate the trained model on the test dataset to assess its performance.

13. Save Model Save the trained model to a specified directory for future use or deployment.

Comparative Analysis:

Table 1: Comparative table for Accuracy

System	Accuracy(%)
Proposed System	99%
Reference [12]	97.5%
Reference [3]	95%
Reference[2]	99%

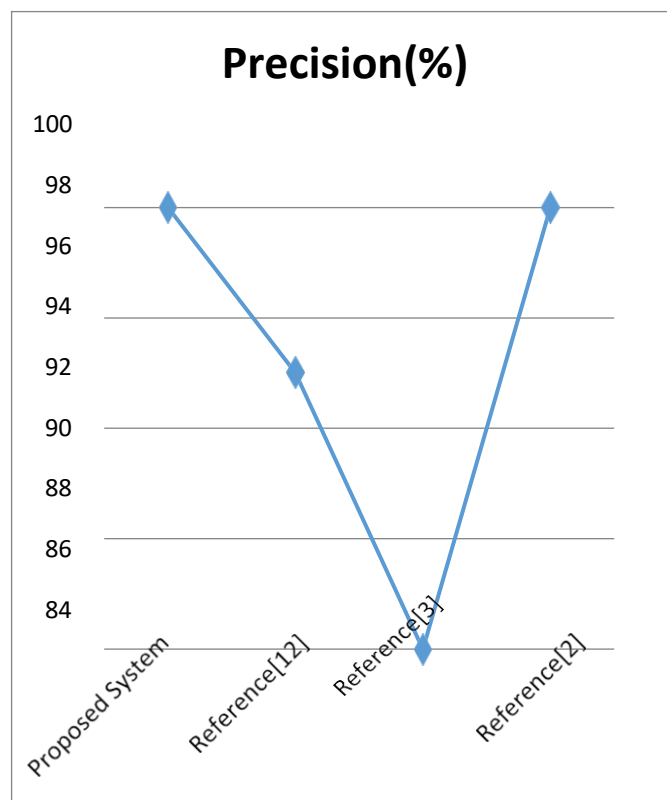


Fig.2. Corresponding graph for table 1 (Dataset = Value/10)

Fig.1. shows the system and Reference[2] demonstrates the highest accuracy (99%) among the systems.

Table 2 : Comparative table for Precision

System	Precision
Proposed System	0.99
Reference [12]	0.9
Reference [3]	0.99
Reference[2]	0.96

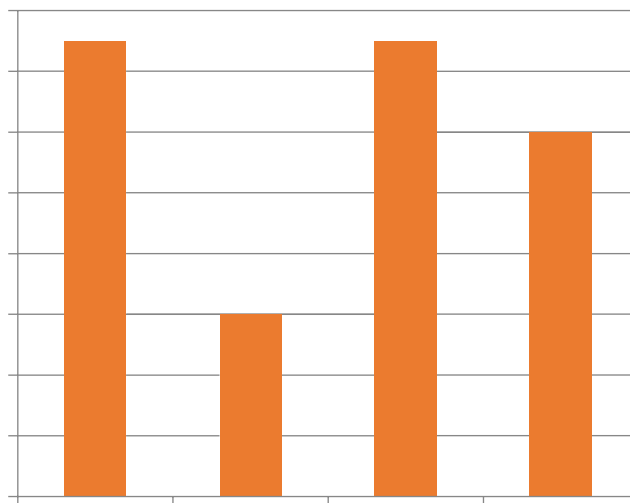


Fig.3. Corresponding graph for table 2(Dataset = Value/10)

Fig.2. shows the proposed system and Reference [3] demonstrates the highest precision (0.99) among the systems.

Table 3 : Comparative table for Recall

System	Recall
Proposed System	0.99
Reference [12]	0.8
Reference [3]	0.99
Reference[2]	0.96

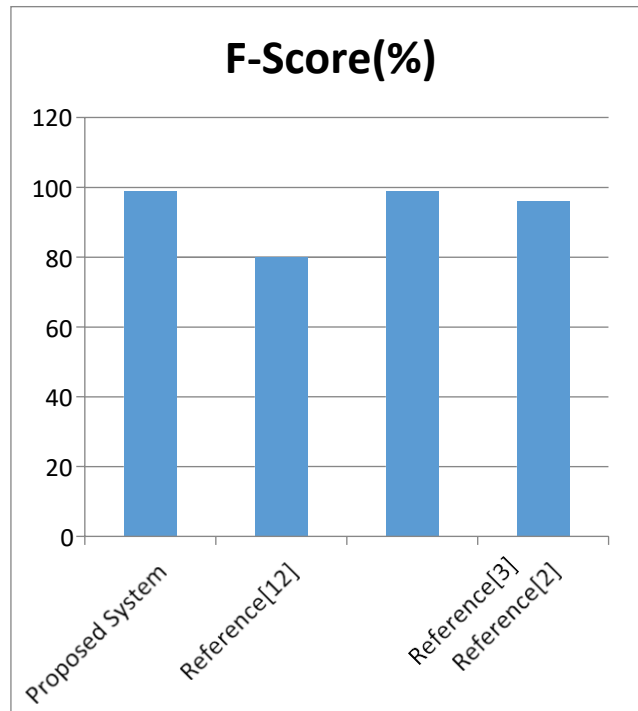


Fig.4. Corresponding graph for table 3(Dataset = Value/10)

Fig.4. shows the table and graph above compare the recall of different Ayurvedic medicine recommendation systems using decision trees.

Table 4: Comparative table for Fscore

System	Fscore
Proposed System	0.99
Reference [12]	0.8
Reference [3]	0.99
Reference[2]	0.96

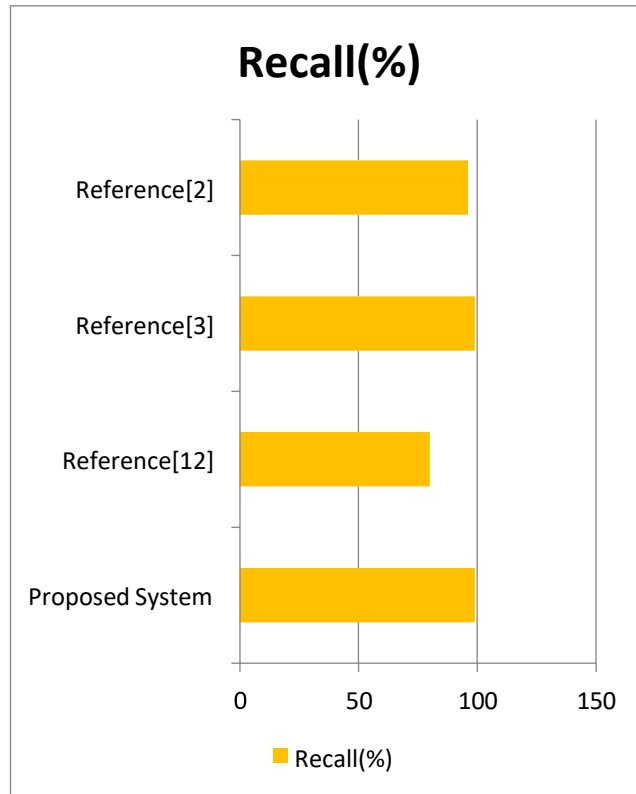


Fig.5. Corresponding graph for table 4(Dataset = Value/10)

Fig.5. shows the table and graph above compare the F -Scoreof different Ayurvedic medicine recommendation systems using decisiontrees.

Tables And Graphs :

Table 5 : Epochs Wise Model Performance

Epoch no.	Accuracy	Loss	Validation Accuracy	Validation Loss
1	0.0768	3.2741	0.2438	2.8507
2	0.3866	2.1638	0.5312	1.3679
3	0.5548	1.5004	0.7595	0.9420
4	0.6830	1.0629	0.7443	0.7981
5	0.7443	0.8297	0.8125	0.5530
6	0.7829	0.6878	0.8438	0.5565
7	0.8339	0.5419	0.8125	0.5312
8	0.8298	0.5182	0.8625	0.4386
9	0.8801	0.3406	0.8562	0.3993
10	0.8739	0.3912	0.8562	0.3993

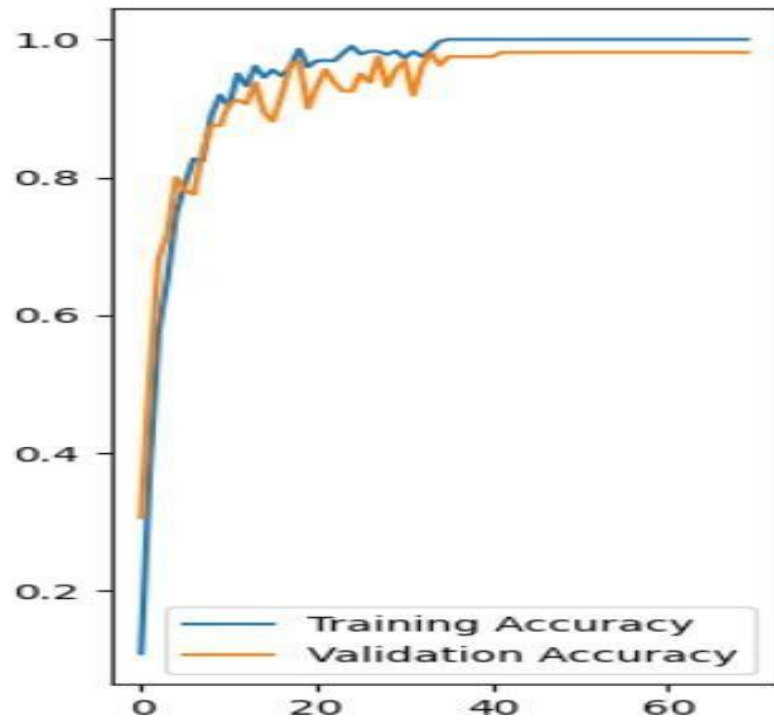


Fig 6 : Training & Validation AccuracyGraph

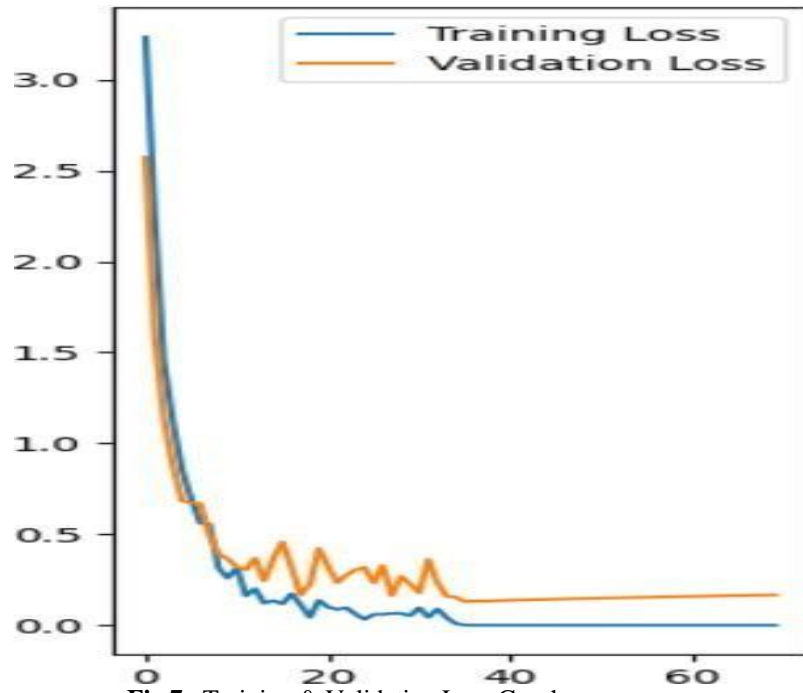






















Fig 7 : Training & Validation Loss Graph

Result :

Sr. No.	Input	Output	Remark
1)		<p data-bbox="564 479 820 517">Actual: Azadirachta Indica (Neem), Predicted: Azadirachta Indica (Neem). Confidence: 100.0%</p> 	<p data-bbox="971 432 1453 562">Actual : Azadirachta (Neem), Predicted : Azadirachta Indica(Neem), Confidence : 100.0%</p>
2)		<p data-bbox="544 837 794 875">Actual: Piper Betle (Betel), Predicted: Piper Betle (Betel). Confidence: 100.0%</p> 	<p data-bbox="971 784 1437 857">Actual : Piper Betle (Betel), Predicted : Piper Betle (Betel),Confidence : 100.0%</p>
3)		<p data-bbox="560 1162 842 1211">Actual: Mangifera Indica (Mango), Predicted: Mangifera Indica (Mango). Confidence: 100.0%</p> 	<p data-bbox="971 1135 1422 1252">Actual : Mangifera Indica (Mango), Predicted : Mangifera Indica (Mango), Confidence : 100.0%</p>
4)		<p data-bbox="549 1518 863 1563">Actual: Pongamia Pinnata (Indian Beech), Predicted: Pongamia Pinnata (Indian Beech). Confidence: 100.0%</p> 	<p data-bbox="971 1487 1445 1733">Actual : Pongamia Pinnata (Indian Beech), Predicted : Pongamia Pinnata (Indian Beech),Confidence : 100.0%</p>

5)		<p>Actual: Syzygium jambos (Rose Apple), Predicted: Syzygium jambos (Rose Apple). Confidence: 100.0%</p> 	<p>Syzygium jambos (RoseApple), Predicted : Syzygium jambos (Rose Apple), Confidence : 100.0%</p>
6)		<p>Actual: Ficus Religiosa (Peepal Tree), Predicted: Ficus Religiosa (Peepal Tree). Confidence: 100.0%</p> 	<p>Actual : Ficus Religiosa (Peepal Tree), Predicted : Ficus Religiosa (Peepal Tree), Confidence : 100.0%</p>
7)		<p>Actual: Plectranthus Amboinicus (Mexican Mint), Predicted: Plectranthus Amboinicus (Mexican Mint). Confidence: 100.0%</p> 	<p>Actual : Plectranthus Amboinicus(Mexican Mint), Predicted : Plectranthus Amboinicus (Mexican Mint), Confidence : 100.0%</p>
8)		<p>Actual: Nyctanthes Arbor-tristis (Parijata), Predicted: Nyctanthes Arbor-tristis (Parijata). Confidence: 100.0%</p> 	<p>Nyctanthes Arbor-tristis(Parijata), Predicted : Nyctanthes Arbor-tristis(Parijata), Confidence : 100.0%</p>

9)		<p>Actual: Hibiscus Rosa-sinensis, Predicted: Hibiscus Rosa-sinensis. Confidence: 99.98%</p> 	<p>Actual : Hibiscus Rosa-sinensis, Predicted : Hibiscus Rosa-sinensis , Confidence : 99.98%</p>
10)		<p>Actual: Carissa Carandas (Karanda), Predicted: Carissa Carandas (Karanda). Confidence: 100.0%</p> 	<p>Actual : Carissa Carandas(Karanda), Predicted : Carissa Carandas(Karanda), Confidence : 100.0%</p>

Scope Of Research:

The developed CNN-based medicinal plant identification system can identify up to 30 different plant groupings with an accuracy of 99%. It establishes a strong basis for subsequent education. Increase the number of plant species in the dataset, adjust it for particular regions, assess and enhance confidence estimates, add additional plant attributes than just leaf photos, incorporate external databases for more comprehensive data, and use dynamic learning continuous user-based improvements based on input could take several forms, such as improving robustness to picture changes, working with botanists to validate and refine datasets, taking ethical and cultural considerations into account, and maximizing scalability and resource use. Together, these routes seek to enhance the system's precision, usability, comprehensiveness, and ethical considerations so that it can be utilized

Future Scope:

Plants with few or no leaves cannot benefit from the suggested methods. For these kinds of plants, there might be attempts to develop classification schemes. A stand-alone single board computer that is linked to a scanner could be used to execute the algorithms. The development of a field- portable system could be possible. Future research on plant identification will use enhanced machine learning classifiers that

incorporate feature selection models and pre-processing to improve accuracy and increase performance.

Conclusion

Plants are essential to human survival. For centuries, indigenous communities have used herbs in particular as folk remedies. Doctors typically use decades of personal sensory or factory experience to identify herbs. Herb identification with scientific data is now simple than ever thanks to recent developments in analytical technologies. This will be helpful to a lot of people, especially those who are not familiar with herb identification. Labor-intensive and requiring proficiency in both sample healing and data interpretation is laboratory based analysis. Thus a simple and reliable technique for identifying plants that mediate conflicts is required. Combining computing and statistical analysis is anticipated to be advantageous for herb identification. The most effective way to quickly identify is with this method. Labor intensive and requiring proficiency in both sample alleviate and data interpretation is laboratory-based analysis. Thus, a simple and reliable technique for identifying plants that mediate conflicts is required. Combining computing and statistical analysis is anticipated to be advantageous for herb identification. This non-destructive approach is the quickest and most accurate way to identify herbs

quickly, especially for those without access to expensive analytical equipment. This study assesses the benefits and drawbacks of various plant identification techniques.

References:

- [1] Venkataraman, Mangayarkarasi N "Support Vector Machine Based Classification of Medicinal Plants Using Leaf Features" published in 2017 International Conference on Advances in Computing, Communications and Informatics (ICACCI)
- [2] Amala Sabu, Sreekumar K, Rahul R Nair "Recognition of Ayurvedic Medicinal Plants from Leaves: A Computer Vision Approach" Published in 2017 Fourth International Conference on Image Information Processing (ICIIP)
- [3] Manojkumar P., Surya C. M., Varun P. Gopi "Identification of Ayurvedic Medicinal Plants by Image Processing of Leaf Samples" Published in 2017 Third international conference on research in computational intelligence and communication network (ICRCICN)
- [4] Vijayashree.T, Dr. A. Gopal "Leaf Identification for the Extraction of Medicinal Qualities Using Image Processing Algorithm" published in 2017 International Conference on Intelligent Computing and Control (I2C2)
- [5] A.Gopal, S. Prudhveeswar Reddy and V. Gayatri "Classification of Selected Medicinal Plants Leaf Using Image Processing " Published in 2012 International Conference on Machine Vision and Image Processing (MVIP)
- [6] Umme Habiba, Md. Rasel Howlader, Md. Aminul Islam, Rahat Hossain Faisal and Md. Mostafijur Rahman"Automatic "Medicinal Plants Classification using Multi-channel Modified Local Gradient Pattern with SVMClassifier" published in Joint 2019 8th International Conference on Informatics, Electronics & Vision (ICIEV)& 3rd International Conference on Imaging, Vision & Pattern Recognition (IVPR)
- [7] R.Janani, A.Gopal "Identification of selected medicinal plant leaves using image feature and ANN" published in 2013 international conference on advanced electronic system (ICAES K. Elissa, "Title of paper if known," unpublished.
- [8] C.Amutha lingeswaran, Mr. M.Sivakumar, Dr. P.Renuga"IDENTIFICATION OF MEDICINAL PLANT'S AND THEIR USAGE BY USING DEEP LEARNING" Published in Proceedings of the Third International Conference on Trends in Electronics and Informatics (ICOEI 2019) IEEE explore Part Number: CFP19J32-ART; ISBN: 978-1-5386-9439-8
- [9] Dileep M.R., Pournami P.N."AyurLeaf: A Deep Learning Approach for Classification of Medicinal Plants" published in TENCON 2019 - 2019 IEEE Region 10 Conference (TENCON)
- [10] Shitala Prasad, Pankaj P Singh"Medicinal Plant Leaf Information Extraction Using Deep Features" published in Proc. of the 2017 IEEE Region 10 Conference (TENCON), Malaysia, November 5-8, 2017
- [11] Sivaranjani.,lekshmi kalinathan, Amutha R. Ruba Soundar kathavarayan, Jagadish kumar.K.J Real-Time "Identification of medicinal plants using machine LearningTechniques" published in Second International Conference on in Computational intelligence in Data science (ICCIDS-2019).
- [12] Ana C. Siravenha, Schubert R. Carvalho, "Plant classification from Leaf Textures" published in 2016 IEEE.
- [13] Prabhat Kumar Thella,"A Brief Review on Methods for Classification of Medical Plant Images "Proceedings of theFifth International Conference on ComputingMethodologies and Communication (ICCMC 2021) IEEE Xplore Part Number: CFP21K25-ART
- [14] Keshav Kumar,"IDENTIFICATION OF AYURVEDIC MEDICINAL LEAVES USING DEEP LEARNING"2021 JETIR August 2021, Volume 8, Issue 8
- [15] S. Roopashree, J. Anitha (2021) - "DeepHerb: Vision Based System for Medicinal Plants Using Xception Features", IEEE Access, VOLUME 9