

**Computer Vision for Plant Identification** 

# Yeni Herdiyeni\* and Julio Adisantoso

Department of Computer Science, Faculty of Mathematics and Natural Sciences,

Bogor Agricultural University, INDONESIA

Tel. +62 251 8625584, Fax. +62-251-8625584, E-mail: <u>veni.herdiveni@ipb.ac.id</u>

\*Contact person and presenter

### Abstract

Regarding to problems in plant identification, Botanist need a tool to assist them in finding the plant taxonomy. There are two goals of plant taxonomy are identification and classification. Computer vision (image understanding) is a discipline that studies how to reconstruct, interpret and understand a 3D scene from its 2D images in term of the properties of the structures present in the scene. We have extensively conducted some research to develop computer vision for plant identification based on image especially for Indonesian medicinal plants and house plant. In the experiment, we used three features for plant identification i.e. morphology, shape and texture. Also, we have conducted some research to classify the plant. There are several classifier algorithms that we used, i.e., Probabilistic Neural Network, Radial Basis Function Neural Network, and Support Vector Machine. To help user finding information easily about the plants, we also conducted research to develop search engine system based on article and research documents. The experiments show that our systens are very promising. In the future we need to explore more extensively mathematical model as function to model the plant in order to improve the accuracy in plant identification and classification.

### I. Introduction

According to ((Damayanti E.K., Hikmat A., and Zuhud, E. 2011) the problems found in identification of medicinal plants are (1) limited number of local people who knows, recognizes, and understands medicinal plants diversity and usage that can identify plants in the field, (2) limited number of plant identification books/guides with many pages in each books/guide which is troublesome to be brought to the field, (3) limited number of authorized institution and facilities for plants identification, (4) limited number of taxonomists in each authorized institution that can scientifically identify enormous plant species, (5) time and money consuming, as effects of problems (1) to (4). They need a tool to assist them in plant identification.

Identification is simply the determination of the similarities or differences between two elements, i.e., two elements are the same or they are different. The comparison of an unknown plant with a named specimen and the determination that the two elements are the same also involves classification, i.e., when one correctly decides that an unknown belongs to the same group (species, genus, family., etc.) as a known specimen, the information stored in classification systems becomes available and applicable to the material at hand (Walters, D.R. & D.J. Keil, ...). Both of processes, identification and classification, involve comparison and judgment and require a definition of criteria of similarities. Plant Identification is determination of identity of unknown plant by comparison with previously collected specimen and plant classification is the process of finding the taxon to which a specimen belongs.

The methods of identification include (1) expert determination, (2) recognition, (3)

2011 International Workshop "Linking Biodiversity and Computer Vision Technology to Enhance

23

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak cipta milik IPB (Institut Pertanian Bogor)

Hak Cipta Dilindungi Undang-Undang

a. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah

**Bogor Agricultural Universit** 



σQ

comparison, and (4) the use of keys and similar devices. In terms of reliability or accuracy the best method of identification is expert determination. Although of great reliability, this method presents problems by requiring the valuable time of experts and creating delays for identification. Recognition is the approaches expert determination in reliability. This is based on extensive, past experience of the identifier with the plant group in question. In some groups this is virtually impossible. A third method is by comparison of an unknown with named specimens, photographs, illustrations or descriptions. Even though this is a reliable method, it may be very time consuming or virtually impossible-due to the lack of suitable materials for comparison. The use of keys or similar devices (synopses, outlines, etc.) is by far the most widely used method and does not require the time, materials, or experience involved in comparison and recognition.

To overcome these problems, we have been conducted research to develop automatically system for plant identification based on image using computer or computer vision for plant identification. Plant identification system is done based on morphology, shape and texture. Plant image that we have used in the experiment is leaf of medicinal plant and house plant.

## II. Computer Vision for Plant Identification

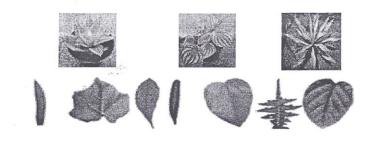
## II.1. Computer Vision

Computer vision is concerned with modeling and replicating human vision using computer software and hardware. It combines knowledge in computer science, electrical engineering, mathematics, physiology, biology and cognitive science. It needs knowledge from all these fields in order to understand and simulate the operation of human vision system.

Computer vision overlaps significantly with the following fields: image processing, pattern recognition and photogrammetry. Image processing focuses on image manipulation to enhance image quality, to restore an image or to compress/decompress an image. Pattern recognition studies various techniques (such as statistical techniques, neural network, support vector machine, etc.) to recognize/classify different patterns. Photogrammetry is concerned with obtaining accurate and reliable measurement from images. It focuses on accurate mensuration.

Computer vision (image understanding) is a discipline that studies how to reconstruct, interpret and understand a 3D scene from its 2D images in term of the properties of the structures present in the scene. The computer vision consists of three levels, i.e.

- 1. Low level vision: process image for feature extraction (edge, corner, or optical flow).
- 2. Intermediate level vision: object recognition and 3D scene interpretation using features obtained from low level vision



3. High level vision: interpretation of evolving information provided by the intermediate level 24

# **Bogor Agricultural Universit**

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak Cipta Dilindungi Undang-Undang

. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah.



<u>0</u>

vision as well as directing what intermediate and low level vision task should be performed. Interpretation may include conceptual description of scene like activity, intention and behavior.

We have been conducted extensive research to develop a computer vision for plant identification automatically. The plant identification system can be used to identify leaf of medicinal plant and house plant. Figure 1 show leaf of medicinal plant and house plant. Images were taken from Botanical Garden Bogor and conservation area at Gunung Leutik Bogor.

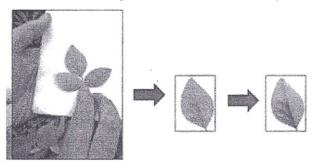


Figure 1. Sample images

### **II.2.** Feature Extraction

Feature extraction is low level vision that providing information for later higher level analysis. It is an essential step in computer vision, performed either separately or jointly with learning process: preprocessing and feature construction are followed by feature selection. Figure 2 show preprocess of image acquisition and image enhancement of leaf. Feature selection is to select the best subset from the input space. Its ultimate goal is to select the optimal features subset that can achieve the highest accuracy results. In our system, we used three features for plant recognition i.e. morphology, texture, and shape.

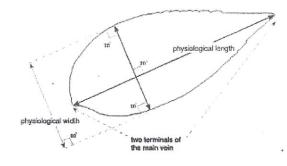


Figure 2. Image selection and enhancement

### Morphology

Plants are basically identified according to their morphological features. We have been observed morphology of the leaf. There are two features of leaf morphology, i.e. basic and derivatives features. The basic features are the diameter, leaf area, and perimeter of leaves. 25

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak cipta milik IPB (Institut Pertanian Bogor)

Hak Cipta Dilindungi Undang-Undang

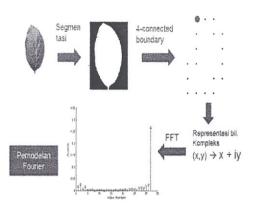
. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah.

Bogor Agricultural Universit



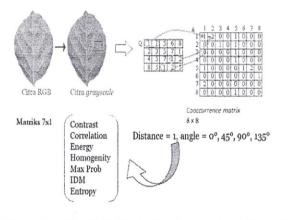
Diameter is the furthest point between the two points from boundary of the leaf. Area is number of pixels that are inside the leaf edge. Perimeter is the number of pixels that are on the edge of the leaf. From three basic features, it can be combined to getting eight derived features such as smooth factor, form factor, rectangularity, perimeter ratio of diameter, perimeter ratio of physiological length and physiological width and vein features.

Figure 3. Physiological length and physiological width



# Shape

Shape of plant leaf is one of the most important features for characterizing various plant species. We have used Fourier transformation to extract the shape of leaf (Rahmadhani, M., 2009). Fourier transformation can provide useful descriptions of shape that independent of position, orientation, and size of an image object. For shape extraction, Fourier transformation needs a binary image. Binary image is used to find set of points in leaf curve. Suppose points (x, y) represented as complex number x + iy, N rows of complex number is called boundary. N



complex numbers is transformed to obtain Fourier descriptor an object. Figure 4 show shape

26

Hak Cipta Dilindungi Undang-Undang

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak cipta milik IPB (Institut Pertanian Bogor)

**Bogor Agricultural Universit** 

a. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah.

2. Dilarang mengumumkan dan memperbanyak sebagian atau seluruh karya tulis ini dalam bentuk apapun tanpa izin IPB b. Pengutipan tidak merugikan kepentingan yang wajar IPB.



extraction using Fourier Transformation.

### Figure 4. Shape extraction using Fourier Transformation

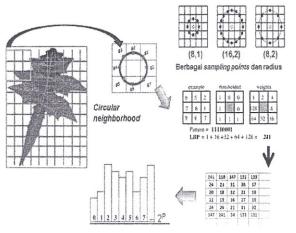
Texture

Co-occurance matrix

Texture analysis plays an important role in many image analysis applications. We have analyzed texture of the leaf using co-occurence matrix (Nurafifah, 2009) and local binary pattern (LBP) (Kulsum, L, 2009). Co-occurrence matrix is a popular representation for texture image using statistical approach. Co-occurrence matrix contain a count of number of times gray level occurs in particular spatial relation to another given features. These matrices well suited for discrimination that are structurally based.

### Figure 5. Co-occurrance Matrix

Figure 5 show co-occurance matrix to extract shape of leaf. The position p is determined by the angle  $\theta$  and distance d. Suppose that in Figure 5, p (1.0), d = 1 and  $\theta$  = 0°, then A (i, j) is expressed by frequency of occurrence along the i-th gray level g(i) at position p with g



(j). Matrix A can be calculated based on the values of texture features. Co-occurrence matrix is computed in four directions consist 0°, 45°, 90°, 135°. Each image will produce four cooccurrence matrix. Seven texture features are obtained based on the co-occurrence matrix, such as energy, inverse different moment, entropy, maximum probability, contrast, homogeneity.

### Local Binary Pattern

The local binary pattern (LBP) operator was developed as a gray-scale invariant pattern measure adding complementary information to the "amount" of texture in images. According to our experiment, LBP has shown excellent performance in many comparative studies, in terms of both speed and discrimination performance (Kulsum L., 2009). The LBP approach is bringing together the separate statistical and structural approaches to texture analysis, opening a door for the analysis of both stochastic microtextures and deterministic macrotextures simultaneously. The LBP operator can be made invariant against rotation, and it also supports

Hak cipta milik IPB (Institut Pertanian Bogor)

# Bogor Agricultural ( Jniversi

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber: 0 ō Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah.

Hak Cipta Dilindungi Undang-Undang

. Pengutipan tidak merugikan kepentingan yang wajar IPB.

2. Dilarang mengumumkan dan memperbanyak sebagian atau seluruh karya tulis ini dalam bentuk apapun tanpa izin IPB



2. Dilarang mengumumkan dan memperbanyak sebagian atau seluruh karya tulis ini dalam bentuk apapun tanpa izin IPB. b. Pengutipan tidak merugikan kepentingan yang wajar IPB.

Ω.

. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah.

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak Cipta Dilindungi Undang-Undang

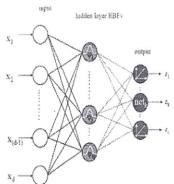
Hak cipta milik IPB (Institut Pertanian Bogor)

**Bogor Agricultural Universit** 

multi-scale analysis.

Figure 6. Local Binary Pattern

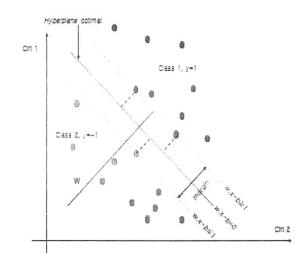
Every pixel in an image generates a single LBP code. Then a histogram is created to collect up the occurrences of different LBP codes from all pixels in the image. This histogram forms the LBP feature vector, which characterize the image texture. To be



different sizes. Defining the local neighborhood as a set of sampling points evenly spaced on a circle centered at the pixel to be labeled allows any radius and number of sampling points.

Decision Class

Decision Layer



**II.3.** Feature Classification

class r

Pattern Layer

Summation Layer

able to deal with textures at different scales, the LBP operator was later

extended to use neighborhoods of

Input Layer

Two goals of plant taxonomy are identification and classification. For plant classification we 28



<u>0</u>

have been extensively conducted research for plant classification using several algorithms. We have used classification several classification algorithm for example Probabilistic Neural Network (PNN), Radial Basis Function Neural Network (RBFNN), and Support Vector Machine (SVM). Structure of PNN is shown in Figure 7. Figure 7. Structure of PNN

PNN consist of four layer namely input layer, pattern layer, summation layer, and decision/output layer. Also we compared performance of Probabilistic Neural Network to Radial Basis Function Neural Network (Figure 8 show structure of RBFNN) and Support Vector Machine (Figure 9 show structure of SVM).

Figure 7. Structure of Radial Basis Function

Figure 8. Structure of Support Vector Machine

# **II.4. Research Results**

Several research results we have conducted are:

1. Feature extraction of morphological and texture of leaf images for leaf image retrieval (Annisa, 2009). This research use three primary morphological features and three

irikatu Ekstraksi							
	A.	$\mathbb{E}$	STALL .	-		and the second s	
1.75	159	RIC	451	J.	D)	D) (	
وصل هل	TSU S	لمصحفال	ill an ar		P		
Sisten Ston	exfilings Da	an dans	en Pro	6.lest	ic No	atal Wotweel	
	- Image of the	Del	1.00	(margers)	9274		
Antie Cite Cites and	onéConunceteùsh rie	othesi rico	nNuraffah-G	640603055	Softwarel	Program UtemaiLeafsh	
		Morfoloci		Tekstur		Bentuk	
identificael.	Dameler	256,204 H	Corbest	0.2852048	1.0231		
4	Perimeter	1319	Cartelation	0.1915500			
	Avea	21556.625	Energy	3 035155			
	Smooth Pactor	1 014861	Homogenety	0.181539			
A MET	Form Factor	0 15555	Max Prob	0.1263098			
	P ratio di D	5.1302114	Mocreed	0.0002976			
WED	Vian Fasture 1	0.0397724	Entropy	0.1259023			
	Ven Feature 2	0.003179;		Carries and the			
Alley Post	Vein Feature 3	0.003289			111	les setti	
	Ven Feature 4	8.026440			0 1111	15 15 20 35	
	Vein Feature 5	8.204870			, ,	Indeks Descriptions	
11	9D6VA						
Nama (Indonesia): Pr							

derivatives morphological features. The primary features are diameter, perimeter, and

29

2. Dilarang mengumumkan dan memperbanyak sebagian atau seluruh karya tulis ini dalam bentuk apapun tanpa izin IPB b. Pengutipan tidak merugikan kepentingan yang wajar IPB.

. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah.

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak Cipta Dilindungi Undang-Undang

Hak cipta milik IPB (Institut Pertanian Bogor)

Bogor Agricultural University



<u>0</u>

area. The derivates features are smooth factor, form factor, ratio of diameter, and five vein features. Co-occurrence matrix is used for texture extraction and Fourier descriptor is used for shape extraction.

2. Shape and vein feature extraction of leaf using Fourier Descriptor and B-Spline Curve (Rahmadhani, M, 2009). In this research, we used Hough Transform and Fourier Descriptor to extract shape and vein of leaf. Experiments result show that leaf shape represented by Fourier Transform descriptor is more effective than represented by Hough Transform.

tirten teng	ierallite for terres	tion.	Bestand
036355	a lakes how many provide a manufacture of a differ	menunana hasi per menunan sa kananan sa s	Samalan Antonio is Sanahan
107500wa kate 300	fataities tijelik ges Dopen Image		Name line. It is any grant determining, had downed date strates and services and
		Extraction Result	fant summe
and the second second	Chosse method	ş .	fabaysian Annus,wang dardij
61	189V - 1 6	÷	Ferninghon - Preceding distances (key of high panels what following personal I technological (Fisher Mg
3	Extent	0	Fungsi - Carassas permenania tantan jain datsam ancara inggual Tantonan menindekan dat kadalah
Query Issage	timik d. 2		Centrepat Centrepatienerse bannaren bilan mangelider, bann dan kunnetse order carro bijelitaria dann (sacht) Emericane noronki, fespati nenets, edit, Stan, kerzina Bingas senzygatigen jenetseki an sekution periodi selamente pati an annan angel nenetse eta sekution sekution periodi selamente anda annan angelitaria eta data eta eta sekutiona sekution periodi sekution periodi sekutiona eta data tenente eta eta sekutiona sekutiona periodi sekution periodi sekutiona eta data tenente eta eta sekutiona sekutiona periodi sekutiona periodi sekutiona eta data eta eta sekutiona sekutiona periodi sekutiona periodi sekutiona eta data tenente eta eta sekutiona sekutiona periodi sekutiona periodi sekutiona eta data tenente eta eta sekutiona eta eta sekutiona periodi sekutiona eta data tenente eta eta sekutiona eta eta sekutiona eta eta sekutiona eta data tenente eta eta eta eta eta data eta eta eta eta eta eta eta eta eta

3. A combine morphological, texture and shape feature based approach for leaf identification using Probabilistic Neural Network (Nurafifah, 2010). This research proposed a new system for leaf identification using classifier combination and Probabilistic Neural Network (PNN). The features that used for classification are morphology, texture, and shape of leaves. This research used co-occurrence matrix and Fourier descriptors to extract texture and shape of leaves respectively. After feature extraction, the feature is classified using Probabilistic Neural Network. Classifiers combination is used to combine these features. The experiment data consist of thirty species flora from Bogor Botanical Garden. Indonesia. The experiment results showed that classification without classifier combination has the accuracy of 79.05%, and using classifier combination the accuracy increased to 83.33%. Hence, the proposed system is promising for leaf identification and supporting plant biodiversity in Indonesia.

Figure 10. Plant leaf identification using combination leaf features

 Automatically plant house identification using Local Binary Patterns Descriptor and 30

# **Bogor Agricultural Universit**

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak Cipta Dilindungi Undang-Undang

. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah



Hak Cipta Dilindungi Undang-Undang

Ω.

Ver Chu Shavak	ital Gebrara	Line I
iii Sharini Angarizani: C () Taki Angari		

Probabilistic Neural Network (Kulsum, L, 2010). This research proposed a new method for a house plant identification using Local Binary Patterns (LBP) descriptor and

idadi Olta B SpinitOx41.pg	(in Panis)	Stored Otra Asti Ot	e Segneklasi	
ercesokan Hengyunalan (1913) (1970-militagir (1971) Ponis Paris (1977) (1973-militagir (1978) Ponis Paris (1977)			Jarak Pagar Jetrophe kones	ina Kiernegaa
escekcekan (18 Jansk Pagar 8) Jansk Pagar	D Keris Nacioj	(i) Kurris Kuping	(gi labal Berb	(h Takat Zanta
56: 0 50: 06/2460 06: 4.20541+29: 06: 04/67/53 CF: 4.20541+20: CF: 0.0054334	505; 0.60307736 545; 0.666417 07; 0.673387	SB: 2.03006 9C: 0.251099 CF: 3.31996	DE: 0.678501	S0: 0.415855 DF: 0.36456 CF: 0.701216
158 128	10	150	150	×

Probabilistic Neural Network (PNN) (Figure 1). There are three kinds of LBP descriptor used in this research, i.e. LBPriu, AR, and LBPV. Database of 300 house plant images belong to 30 different types of house plant in Indonesia are extracted using LBPriu, VAR, and LBPV based on texture feature and classified using PNN. The experimental result showed that LBPV has the best accuracy to indentify house plants with accuracy 73.33%. The proposed system is promising because system is capable to identify house plants species efficiently and accurately.

Figure 11. Automatically plant house identification using Local Binary Patterns Descriptor and Probabilistic Neural Network

31

Hak cipta milik IPB (Institut Pertanian Bogor)

**Bogor Agricultural Universit** 

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah.

2. Dilarang mengumumkan dan memperbanyak sebagian atau seluruh karya tulis ini dalam bentuk apapun tanpa izin IPB b. Pengutipan tidak merugikan kepentingan yang wajar IPB.



σΩ

5. House plant image segmentation using Boycov Method and Kolmogorov Max Flow/Min Cut graph (Wibowo, R. W, 2010). The objective of this research is to remove the complex background of image using Boycov Method and Kolmogorov Max Flow/Min Cut graph. Segmentation is used to increase the accuracy of identification especially for image with the complex background.

6. Leaf Matching using B-Spline Curve (Hikmawati, I. N, 2010). This research proposed an approach that incorporates deformable mapping (deformation energy) and geometric characteristics (strain difference) of spline curve to measure leaf similarity. This method was implemented on leaf image (Figure 3). Experiment result showed that same species leaves have low total cost as compared to different species leaves.

# Figure 12. House plant image segmentation using Boycov Method and Kolmogorov Max Flow/Min Cut graph

Figure 13. Leaf matching using B-Spline curve



- Medicinal Plant Identification Based on Morphology, Shape and Texture Using Probabilistic Neural Network (Nurfadillah, E., 2011). This research proposed a new system for Indonesian medicinal plant identification. The experiments result show that combination of plant features can improve the accuracy of plant identification until 74.67%.
- 8. Classifier Combination for Leaf Identification (Heriningtyas, P., 2011). This research compares a several classifier algorithms for plant leaf classification. We compared three classifiers i.e. Probabilistic Neural Network, Radial Basis Function Neural Network and Support Vector Machine. The Experiments result show that classifier Probabilistic Neural Network is superior to classify plan leaf image for identification.

### II.5. Search Engine

Some plants for medicinal use have been documented in the form of books, reports and papers by researchers, students, and practitioners. With increasing number of research on medicinal plants, the number of books, reports, and papers will be increasing. This will make searching 32

Hak cipta milik IPB (Institut Pertanian Bogor)

**Bogor Agricultural Universit** 

. Pengutipan tidak merugikan kepentingan yang wajar IPB.

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak Cipta Dilindungi Undang-Undang

Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah.



information is not easy. So it is need to store and retrieve written information using computer. Information Retrieval (IR) is finding material (usually documents) of an unstructured nature (usually text) that satisfies an information need from within large collections (usually stored on computers).

The field of information retrieval also covers supporting users in browsing or filtering document collections or further processing a set of retrieved documents. Given a set of documents, clustering is the task of coming up with a good grouping of the documents based on their contents. It is similar to arranging books on a bookshelf according to their topic. Given a set of topics, standing information needs, or other categories (such as suitability of texts for different age groups), classification is the task of deciding which class(es), if any, each of a set of documents belongs to. It is often approached by first manually classifying some documents and then hoping to be able to classify new documents automatically.

We have been conducted some research in information retrieval, especially for document indexing, classification, clustering, and question answering system (Paskianti, K, 2011; Ramadhina, A., 2011; Indriyani, W., 2011, Herawan, Y., 2011, Pramujadi, A, 2009).. In term of computer vision technology for plant identification, our research in information retrieval is very important to support identification and to provide informations. Figure 1 show information retrieval for medicinal plant document.

Figure 14. Information retrieval system for medicinal plant

### **III.** Conclusion

Computer vision (image understanding) is a discipline that studies how to reconstruct, interpret and understand a 3D scene from its 2D images in term of the properties of the structures present in the scene. We have extensively conducted some research to develop computer vision for plant identification based on image especially for Indonesian medicinal plants and house plant. In the experiment, we used three features for plant identification i.e. morphology, shape and texture. Also, we explored several classifier to classify the plant. The classifier that we used, i.e., Probabilistic Neural Network, Radial Basis Function Neural Network, and Support Vector Machine. To help user finding information easily about the plants, we also conducted research to develop search engine system based on article and research documents. The experiments show that our systems are very promising. In the future we need to explore more extensively mathematical model as function to model the plant in order to improve the accuracy in plant identification and classification.

### REFERENCES

- Annisa. 2009. Morphology and Texture Feature Extraction For Leaf Image retrieval. Skripsi. Department of Computer Science, Faculty Mathematics and Natural Sciences. Bogor Agricultural University.
- Herawan, Y. 2011. Feature Extraction of Medicinal Plants Using Chi-Square with Naïve Bayes Classifier. Skripsi. Department of Computer Science, Faculty Mathematics and Natural Sciences. Bogor Agricultural University. Unpublished
- Heriningtyas. 2011. A Classifier Combination for Plant Leaf Identification. Skripsi. Department of Computer Science, Faculty Mathematics and Natural Sciences. Bogor Agricultural University. Unpublished
- Hikmawati, I., N., 2010. Leaf Matching Using B-Spline Curve.. Skripsi. Department of Computer Science, Faculty Mathematics and Natural Sciences. Bogor Agricultural University.
- Indrivani, W., 2011. Pseudo-Relevance Feedback for Information Retrieval using Sentence

33

2. Dilarang mengumumkan dan memperbanyak sebagian atau seluruh karya tulis ini dalam bentuk apapun tanpa izin IPB b. Pengutipan tidak merugikan kepentingan yang wajar IPB.

Dilarang mengutip sebagian atau seluruh karya tulis ini tanpa mencantumkan dan menyebutkan sumber:

Hak cipta milik IPB (Institut Pertanian Bogor)

Hak Cipta Dilindungi Undang-Undang

a. Pengutipan hanya untuk kepentingan pendidikan, penelitian, penulisan karya ilmiah, penyusunan laporan, penulisan kritik atau tinjauan suatu masalah

**Bogor Agricultural University**