# وقائع المؤتمر العلمي السنوي الثاني والعشرون لقسما الحاسبات والعلوم / كليت التربيت الاساسيت/ الجامعة المستنصرية والموسوم (البحث العلمي ركيزة التنمية المستدامة) 8\_9 آيار 2022 وتحت شعار (البحث العلمي بوابتنا للبناء والتقدم)

#### A Survey of Image Segmentation Based on Soft Computing Approaches Saja Hikmat Dawood

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#### **Abstract**

Image segmentation is an essential step in image processing, that aim to make image analyzing easier. Image segmentation can be done by grouping or partitioning the image pixels into identical sets (regions) by relying on some qualities such as color, or texture, etc. It's have been used in many field such as object detection, recognition tasks, medical imaging and much more, there are basically two primary approaches for image segmentation which are traditional approaches and soft computing approaches (SCA), a lot of methods have been proposed based on these two approaches. SCA have many advantage over traditional approaches like flexibility, cost-effective, high performance. SCA involve using fuzzy logic, Artificial Neural Network (ANN) and Genetic Algorithm. This paper focus on providing a state-of-the-art new review and summaries of researchers' work on image segmentation based on different SCA. That will help the new researchers to learn about these methods and then choosing a certain method from these for improving or developing it to produce a new method for image segmentation.

**Keywords:** - Image Segmentation, Soft computing, Fuzzy logic, Neural network, Genetic algorithm

#### I. Introduction

An Image can store a lot of useful information. A digital image composed of a finite number of elements (these elements also known as pixels or picture/ image elements), each element has a particular value and location. pixels are the smallest single element in the image, retaining a finite, discrete, quantized values that represent the brightness, intensity or gray level at any specific point [1]. Understanding the image and extracting data from it to do specific tasks is a key area of application for digital image technology, and image segmentation is the first stage in this process.

Image segmentation considered as one of the fundamental steps in image processing [2], it attempts to automatically analyze or interpret an image. Segmentation fill the gap between low-level and high-level image processing [3], as shown in Figure 1. So image segmentation can be defined as the

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process of dividing an image into several segments based on particular characteristics including color, texture, and intensity values etc, and that's done in order to transform the image's representation into something more meaningful and easier to understand (similar in term of properties, attributes and features). It genuinely locates an image's objects and bounds. The degree to which this division is carried out is determined by the problem being solved; segmentation should finish once the application's objects of interest have been isolated. So Segmentation is a technique for converting a complicated image into a simpler image [4] [5].

There are many applications of Image segmentation, such as medical imaging (medical diagnosis, locate tumors and other pathologies, etc.), content based image retrieval, Recognition tasks(recognition of Face, fingerprint, Iris), Machine vision, industrial inspection, object detection (like detecting roads and bridges in satellite images, Face detection, etc.), and more [5].

Image segmentation can be accomplished using a variety of approaches. Traditional approaches and soft computing approaches. Traditional approaches are straightforward and easy to implement methods. They achieve and produce precise answers to the segmentation problem. Traditional approaches are usually classified into four categories based on how they work: (1) region-based segmentation, (2) clustering, (3) edge-based segmentation, and (4) thresholding. These approaches are inefficient because they cannot handle real-world complicated situations that tolerate partial truth, imprecision, uncertainty, and approximations; they are also time-consuming, inefficient, and labor-intensive.

As a result, Soft Computing approaches(SCA) are utilized to address such issues. Soft Computing's objective is to create artificial intelligence by replicating a human brain's reasoning capabilities to solve ambiguity or real-world complicated situations. Soft computing(SC) combines computing techniques with biological structures to create new methods for more dynamic, competent, and dependable solutions. SC is a collection of techniques that includes fuzzy logic (FL), artificial neural networks (ANN), and genetic algorithms (GA). SC approaches, unlike traditional approaches, are forgiving of imprecision, uncertainty, partial truth, and approximations. Working with their ownership function gives them greater freedom, which makes them more powerful. SCA are widely employed and valued by researchers because of their adaptability and accuracy. SCA have the added benefit of being cost-effective, high-performing, and dependable solutions to

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difficult issues. SC techniques have been used in a variety of fields, including scientific study, medicine, engineering, and management.

So, the contribution of this paper is to provides review and summaries of researchers' work on image segmentation based on different soft computing approaches (SCA). The goal is to bring together various aspects such as new techniques, emerging topics for segmentation depending on SCA, evaluation parameters, problems associated with it, future plans, and other resources so that researchers can focus on developing new methods for various image segmentation applications [6].

This paper is organized as follows, Section II, is focused on Review some of the new works in image segmentation based on different soft computing approaches. Section III presents the conclusion for soft computing approaches to image segmentation. Finally, the references are given in Section IV.

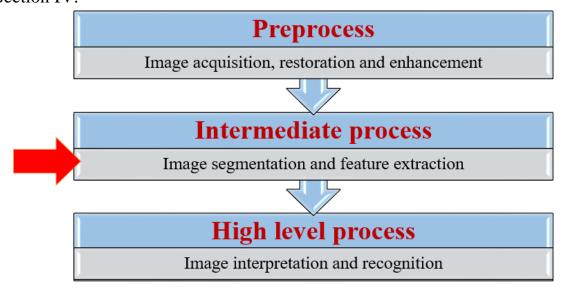


Figure 1: Element of image Analysis

#### **II. Soft Computing Approaches**

For comparison, three different soft computing approaches to image segmentation are most commonly used. These are (1) Fuzzy based approaches, (2) Genetic Algorithm based approaches, and (3) Neural Network based approaches. The following are the specifics of Approaches:

#### 1. Fuzzy based Approach

It is based on the degree of truth or falsehood principle. It's a probabilistic segmentation method based on a fuzzy logic framework. The input image is first fuzzified, then passed on for membership modification based on expert knowledge and a set of fuzzy logics and set theory. The final result is

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achieved after passing the modified image through image defuzzification, Figure 2, show the general structure of fuzzy system. Fuzzy thresholding, fuzzy integral-based decision making, and fuzzy c-means clustering are examples of fuzzy-based approaches [7]. The summary of fuzzy logic(FL) newest articles are shown in Table 1.

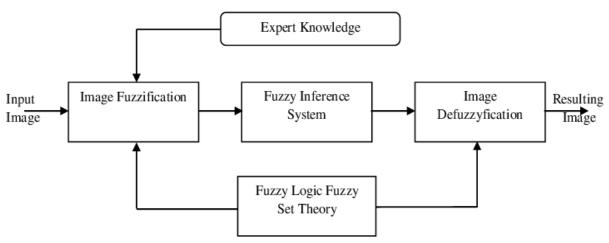


Figure 2: General structure of fuzzy system

Table 1: summary of FL newest articles

R	TECHNI	Parameter	Domain	Data set/
ef	QUE/			Result
	methodol			
	ogy			
[8	Fuzzy	$n_{eb} = n_{ob} = 50.K = 100.M = 100$	Binary	Various
]	Entropy		and	images/
	and		Grayscale	six
	improved		images	functions
	Intelligen		_	are used
	t			to test
	optimizat			reliabilit
	ion			y of
	algorith			chaos
	m			bee
				colony
				particle
				swarm
				optimiza
				tion
				(PSO)

				and PSO
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				value
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				standard
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				also
				used.
				The
				proposed
				algorith
				m has
				good
				perform
				ance in
				aspect of
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	A	1 1 1 1 1 1	CARA A	stability.
[9	Active	$\lambda_1 = \lambda_2 = \mu = 1. \kappa = 2. \rho = 3.\Delta t = 0.1.u_1 = v_1 =$		$0SXOR \times (255)$
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				n. The
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				e offers
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				y and accuracy compare d with the RSF model
[1 0]	Fuzzy Clusterin g	Iteration (100, 200, 300 and 400)	brain tumour image (MRI images)	global database s of brain tissue images (MRI images)/ the proposed algorith m techniqu e correctly (accurate ) detecting brain tumor. it can distingui sh the brain tissue of the three regions (WM, WG and CSF).
[1 1]	Fuzzy Set and Threshol ding	image gradient (High Pass Filter and mean filter $(hpf\ and\ mf)$ ) and thresholding filter	Mammog raphy Images Segmenta tion	database (mini- MIAS)/ the proposed method produce

The proof of th
time consumi ng to impleme

				method.
Г1	itorotivo	m -1.1 m -5.0	MRI	medical
[1 3]	iterative type-2	$m_1$ =1.1, $m_2$ =5.0	images on	images
ال	fuzzy C-		a a	from a
	mean		massively	large
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	(1121 C M)		SIMD	/
	141)		architectu	
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				FCM.
				but it is
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				terms of
				processin
				g time.
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				proposed
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				ntly
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				n time
				over the
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				accuracy
				and
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				n time
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#### 2. Genetic Algorithm based Approaches

A genetic algorithm is a method for computing, it is an algorithm for searching optimal solutions. by modeling the biological evolution process through natural selection and genetic mechanisms. It creates individual populations, determines individual fitness, and selects regenerative individuals, crosses, and mutates to create new individuals based on fitness. It features a high level of robustness, parallelism, adaptability, and speed of convergence. It can be used to determine the segmentation threshold in image segmentation technology [14]. Image segmentation using genetic algorithm shown in Figure 2. The summary of genetic algorithm (GA) newest articles are shown in Table 2.

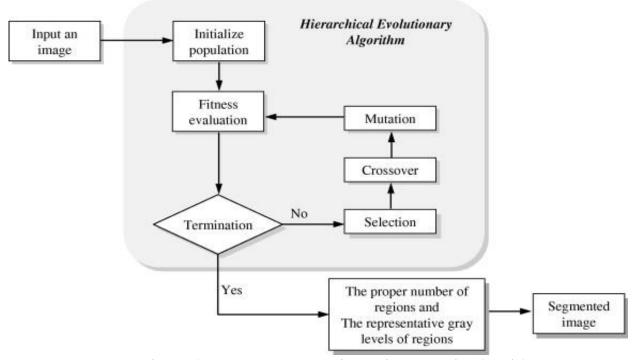


Figure 2: Image segmentation using genetic algorithm

and here is a review of some newest previous studies for this method.

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**Table 2:** summary of GA newest articles

Ref	TECHNIQ	Parameter	Dom	Data set/	FUTURE
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	gy				
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	OTSU	10	image	The improved	method
	method	populations,	s	Otsu approach	
	based on	optimal		can reduce	
	Genetic	threshold		calculation	
	Algorithm	range $[T1,$		time in the	
		T2].		long run	
		individual			
		fitness:			
		$\sigma 2(kk)$			
		<b>Mutation:-</b>			
		Hamming			
		distance $H_{i,j}$			
		length of the			
		binary			
		number $n =$			
		8.			
		mutation rate			
		$P_m$			
		$f.f_m.f_a.H_a.H_m$			
F4 63	3.5	<b>-</b>	0 .	11.00	-
[16]	Maximum	Initializatio	foreig	different	Improving
	Entropy	n: -	n C'1	foreign fiber	the
	and	20	fiber	images / The	accuracy
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	Algorithm	max.no.	cotto	algorithm is	proposed
		iteration=10	n ·	more accurate	algorithm
		0	image	than the	for
		probability	S	standard Otsu	detection

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length image histograms on a	nd
m = 50 s and sizes / edge	
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condition is RMSE are by	
based on the used to consider	n
maximal evaluate the g oth	er
number of suggested entropies	
generations method which and a lar	ge
which set to maximizes sample	_
300. entropy (by real- wo	ld
using Tsallis and	

produces better image segmentation quality than the standard thresholding strategy. The result show that Renyi entropy provides better image segmentation quality.  Improved Max.no. Fuzzy C- iterations $T$ , termination Based on threshold $\varepsilon$ and treshold $\varepsilon$ and Genetic $\varepsilon$					· ·	-
					_	images
					-	
[18]       Improved Fuzzy C-Means Based on Genetic Algorithm       Max.no. iterations $T$ , termination threshold $\varepsilon$ and $T$ real membership S-FCM algorithm       synth etic, simulated, and actual medical images / The results suggest that the robustness to noise of the new image segmentation is significantly improved by GA.         [18]       Improved Fuzzy C-Means Genetic all of Genetic					better image	
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[18] Improved Max.no. synth synthetic, hetic, simulated, and simul actual medical images / The Genetic $=100$ and results suggest that the weighted index $m=2$ al noise of the image segmentation algorithm suppression parameter $\alpha$ $=0.9$ , GFCM convergence speed parameter $\alpha$					better image	
[18] Improved Fuzzy C- iterations $T$ , etic, simulated, and Means Based on Genetic Algorithm  For iterations $T$ , etic, simulated, and actual medical images / The results suggest that the weighted index $m = 2$ al membership image S-FCM algorithm suppression parameter $\alpha = 0.9$ , GFCM convergence speed parameter $\alpha$					_	
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Means Based on threshold $\varepsilon$ Genetic Algorithm $10^{-5}$ , weighted index $m = 2$ membership S-FCM algorithm  suppression parameter $\alpha$ =0.9, GFCM convergence speed parameter $\alpha$	[18]	_				Nil
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$\begin{array}{c ccccccccccccccccccccccccccccccccccc$					, ,	
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convergence speed parameter α			<del>*</del>		GA.	
speed parameter α			·			
parameter α			_			
			speed			
=0.9, PFCM						
algorithm			algorithm			
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=1, b			=1, b			

		=1, η=2.			
[14]	Genetic	The object is	Palm	Palm (256 *	Nil
	Algorithm	Palm	(256	256) image /	
		(256*256)	*	the genetic	
		image, the	. ′	algorithm is	
		initial	image		
		population	•	optimization	
		number is set		algorithm in	
		to 20, the		image	
		crossover		segmentation,	
		probability		which reduces	
		$P_c$ is 0.8, the		the time	
		mutation		required to	
		probability		find the	
		$P_m$ is 0.03,		threshold. The	
		and the		effect of	
		maximum evolution		image	
				segmentation using a	
		algebra T is 100.		genetic a	
		100.		algorithm is	
				superior to	
				standard	
				image	
				segmentation,	
				it also offers	
				strong	
				stability and	
				enhances	
				convergence	
				speed.	
[19]	Hybrid	length of	practi	various image	Nil
	Genetic	chromosome	cal	kinds / PSNR	
	Algorithm	m = 50, size	and	and MSE are	
	with	of population	medic	used for result	
	Particle	p = 30, max	al	validation. the	
	Swarm	generation	image	proposed	

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	Optimizati on and Entropy (GA-PSO and Renyi, Tsallis entropies)	thresholding rate pairs (t*,	sampl es	method proficiently increases entropy and produces improved image segmentation quality when compared to the conventional	
				thresholding method	
[20]	modified Genetic Algorithm (GA) by k- means algorithm	Nil	cell blood image s from micro scope	set of cell blood images /	Nil

#### 3. Neural Network based Approaches

The nervous system of a human processed information in the same way as a "Artificial neural network (ANN)" did during image processing operations. The weights between the layers of the network, also known as interconnectivity, are changed so that the network can learn or understand the knowledge. When the network has fully grasped or been trained on the knowledge, it is capable of determining the best solution output for the set of input data. Generalization is the most significant property of neural networks [21]. The architecture of artificial neural network illustrated in Figure 2. The summary of neural network (NN) newest articles are shown in Table 3.

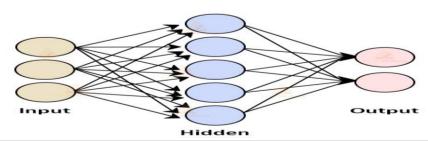


Figure 3: Architecture of Artificial Neural Network

**Table 3:** summary of NN newest articles

Ref	TECHNI QUE/ methodol ogy	Domain brain	No. of convoluti onal layer	No. of Maxpool ing/ average pooling/ sample	Filly connec ted layer	Softm ax classif ier/ outpu t layer	Type of classifica tion rectifier/ cost function/ activatio n function Rectifier	Trainin g using BPA or SGD/ other parame ters	Data set/ Result/ Future work  datasets BRATS
	on neural network (CNNs)	tumor image segment ation	(18*28) 2C (12*12) 1C (5*5)	(28*28) 2S (12*12) 1S (5*5)	(28*28 ) 1F (12*12 ) 1F (5*5)	outpu t layer	linear units (ReLU)	forwar d and backwa rd propag ation (BPA)	2013 / Dice ratio (DR) is used for evaluation the proposed method, CNNs accuracy is better than other methods like (Bauer and Menze).
[23]	Convoluti onal Neural Networks	brain tumor (gliomas ) segment ation from MRI images	3C (3*3*2) 22 deep layers	average pooling	3F	1 outpu t layer	Activatio n function: Leaky ReLU (LReLU)  In the last layer: Softmax  / Batch normaliz ation / cross entropy as loss function	Stochas tic gradien t descent (SGD)	BRATS 2015 databases / the performance of whole, core, and enhancing areas evaluated using Dice =0.84, 0.79, 0.75, Positive PPV= 0.88, 0.86, 0.70, and Sensitivity = 0.82, 0.75, 0.86. training duration =140m. the proposed strategy is efficient and time-saving /segmentation the lung nodule depending on the outcomes of this study.
[24]	Deep Neural Networks using the U-net architect ur	Lung CT Image Segment ation	2C (3*3)	2P (2*2)	-	-	Rectifier linear units (ReLU)	-	Lung Image Database Consortium image collection (LIDC- IDRI) / accurate segmentation with 0.9502 Dice- Coefficient index was obtained

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### Table 3: continued

			1 a	Die 3: C	Ontini	icu			
Ref	TECHNI	Domain	No. of	No. of	Filly	Softm	Type of	Trainin	Data set/ Result/
	QUE/		convoluti	Maxpool	connec	ax	classifica	g using	Future work
	methodol		onal layer	ing/	ted	classif	tion	BPA or	_ *************************************
	ogy		Jim injei	average	layer	ier/	rectifier/	SGD/	
	ogy				layer			other	
				pooling/		outpu	cost		
				sample		t	function/	parame	
						layer	activatio	ters	
							n		
							function		
[25]	Deep	Medical	• Enc	1P	-	1	Rectifier	-	dataset is from
	Multiscal	Image	oder			Softm	linear		ADNI
	e	Segment	:-			ax	units		(Alzheimer's
	Convoluti	ation	3C layers				(ReLU)/		Disease
	onal		$(1 \times 1, 3 \times$				batch		Neuroimaging
	Neural		$3,5\times 5$				normaliz		Initiative) /
	Network		• U-				ation		evaluation indexes
	Model		_				(BN)/		are DSC= 92.54,
			Net						
	(DMCNN		Mod				sigmoid /		SEN=91.87, and
	)		el:				loss		PPV=92.08,
			28C layer			1	function		Time=8.5s. the
			end of U-						new strategy
			net:						decreases
			1C layer						computation time
			$(1 \times 1)$						and improves
			• Dec						segmentation
			odin						accuracy. It also
			g						offers a high level
			e Part						of robustness
			:-						when compared to
									other
			1C layer						segmentation
									methods.
									future work is to
									segment different
									types of images
									using deep
									learning methods
[26]	deep	Alzheim	(3*3)	(2*2)	-	3	loss	-	OASIS, AD-86
	learning	er's				tissue	function		and AD-126
	in	diagnosi				S			datasets / for
	segmenti	s in 3D				segme			segmentation in
	ng and	brain				ntatio			both datasets
	classifyin	MR				n			Dice= 0.96 and
	g	images				groun			accuracies 0.88,
	8					d			and 0.80 for
						truth			classification,
						սաս			
									respectively. \ The
									proposed
									approach was
						1			tested and found
									to be quite
									accurate. \ To
									improve CNN
									model, some
									alternatives have
									been proposed,
									such as
									segmenting more
									brain tissues or
						1			lowering loss
									_
									information in downsampling
	<u> </u>				l	1	l		uownsampning

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#### **III. Conclusion**

In image processing, Segmentation is an important pre-processing step in the areas of image analysis. It is a critical and essential component of image recognition system and usually determines the quality of the final result. Different approaches have been introducing, but the accuracy is the main problem for all of these. Therefore, the use of soft computing approaches (SCA) was adopted. SCA involves, fuzzy based approach, genetic algorithm based approach, and neural network based approach. the contribution of this article's is to providing new reviews and summaries for some of the researchers' work on image segmentation based on soft computing approaches (SCA). these approaches were tested on a real-life images, the hopes that it will encourage academics to work on inventing new segmentation methods with more accurate results.

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#### دراسة استقصائية لتجزئة الصور بالاعتماد على طرق الحوسبة اللينة

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#### مستخلص البحث:

تعد تجزئة الصور خطوة أساسية في معالجة الصور، والتي تهدف إلى تسهيل تحليل الصور. يمكن إجراء تجزئة الصورة عن طريق تجميع أو تقسيم بكسلات الصورة إلى مجموعات (مناطق) متطابقة من خلال الاعتماد على بعض الصفات مثل اللون ، أو الملمس ، وما إلى ذلك. وقد تم استخدامها في العديد من المجالات مثل اكتشاف الأشياء، ومهام التعرف، والتصوير الطبي وفي الكثير من المجالات الاخرى. هناك طريقتان أساسيتان لتجزئة الصور وهما الطرق التقليدية وطرق الحوسبة اللينة بالعديد من SCA ، وقد تم اقتراح الكثير من الطرق بناءً على هذين النهجين. تتمتع طرق (SCA) المزايا التي تفوق الطرق التقليدية مثل المرونة والفعالية من حيث التكلفة والأداء العالي. تتضمن طرق والخوار زمية الجينية. تركز (ANN) استخدام المنطق الضبابي والشبكة العصبية الاصطناعية SCA والخوار وقة البحثية على تجزئة الصور المختلفة. سيساعد ذلك الباحثين الجدد في التعرف على هذه الأساليب ثم اختيار SCA بناءً على طرق المختلفة. سيساعد ذلك الباحثين الجدد في التعرف على هذه الأساليب ثم اختيار SCA بناءً على طرق المختلفة.

الكلمات المفتاحية: - تجزئة الصورة، الحوسبة اللينة، المنطق الضبابي، الشبكة العصبية، الخوار زمية الجبنية.