A Review on Image Segmentation Technique

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ABSTRACT: Making a segment of an image or any object is what segmentation is all about. The first stages in image segmentation are pattern recognition and picture analysis. We may conduct significant research topics in the segmentation of video with dynamic backdrop in the computer vision area and image analysis. In image processing and analysis, picture segmentation is the most common judging or analyzing function. Picture segmentation is the division of an image into distinct areas that are same or similar in some aspects and homogeneous in others. The outcomes of picture segmentation have an impact on image greater tasks and evaluation Characterization and depiction of objects, as well as feature measurement, are all part of image analysis. The categorization of objects is followed by a higher-order job. As a result, image segmentation relies heavily on characterization, visualization of the part of attention in any image, and demarcation. The current methods of picture segmentation are examined using various algorithms in order to allow user interaction with images. The review of picture segmentation is presented in this article utilizing several methods.

KEYWORDS: Image Segmentation, Image Analysis, Segment, Video.

I. INTRODUCTION

Video image segmentation is used to create a dynamic background. Video segmentation with a dynamic background has long been a hot topic in intelligent surveillance and human-machine interface research[1]. Images are required for segmentation. However, The photos are either monochrome or colorful. The grey level causes color images. The color of the color image changes as the grey level contrast changes. Medical image segmentation relies heavily on image segmentation[2]. Biomedical photographs serve a critical function in supporting health-care professionals, allowing patients to receive treatment. As a first step in Medical Image Analysis, segmentation is critical for medical images (MIA). Image measurement, image display, and feature extraction are all examples of errors in image analysis. As a result, proper image segmentation in medical images is difficult due to disparities in head, leg, and brain components sizes, illness kind, and other factors.

Manuscript Received November 19, 2020

Bharti Sharma, Assistant Professor, Department of Computer Application, Vivekananda Global University, Jaipur, India (Email Id- bharti.sharma@vgu.ac.in) As a result, different algorithms and procedures are required for the segmentation and classification of medical images. However, dependent on the radiologist's background, he may devote time to studying medical pictures that need interpreting. To overcome these limitations, the use of computer-aided systems becomes essential [3]–[5].

Digital image processing is an artificial intelligence method that, when when fuzzy logic, pattern identification, and deep training are mixed, is extremely useful. Image techniques may be classified into one of two categories: image engineering and image processing (IE)[6]. The top layer of image architecture is used for image understanding, the mid level for image treatment, and the lowest layer for image processing[7]. The earliest and least difficult step of image processing is picture separation, with the goal of extracting information from images that is represented as data through image segmentation, feature measurement, and object representation[8]. The precision of feature measurement has a significant impact on the picture segmentation outcome. Image segmentation is a computer-assisted process, and computerization of medical image segmentation is crucial in medical imaging applications. The procedure of dividing an copy into its constituent pieces and extracting the portions of interest or objects is known as image segmentation[9]. Automatic picture segmentation is also performed, but the most important job is that the segmentation results influence all future image analysis procedures. Image (and video) segmentation is an important phase in image analysis that occurs in the intermediate layer of image engineering, meaning it is affected by both data and human variables. Image analysis includes activities like feature measuring, object representation and characterization, and potentially elevated tasks like object classification and scene comprehension [10], [11]. Preprocessing, separation of the source picture, photo removal and selections, identification, and comment are the five stages of automated medical image segmentation by intelligent technique, with the second and third steps combining to form the process of photo analysis.

They are classified into two kinds based on picture segmentation assessment techniques:

- 1. Identification
- 2. Compare and contrast

Characterization is an intra-technique process, while comparison is an inter-technique one. Picture segmentation methods are presently classified into the following groups based on two image characteristics and various technology[12].

A. Detecting Discontinuities

The identifying discontinuities feature, which comprises an image segmentation method, is required for edge detection[13]. The picture's intensity is altered, and the image is partitioned. Edge detection is a method of segmentation that involves locating pixels on the region's edge. The border between neighboring portions of a picture may be characterized as an edge.

B. Detecting Similarities

Picture segmentation techniques such as thresholding, region expanding, region splitting, and merging are used to divide an image into comparable areas based on a set of preset criteria. Thresholding is a popular method for region-based segmented, in which a photo is portrayed as a collection of dots with numbers greater than or equivalent to the criterion and numbers less than the cutoff[14]. Thresholding may be employed in situations when the user wishes to focus on the most important details or parts of a picture. Clustering is a method for segmenting regions in which a picture is divided into sets or clusters of pixels with comparable feature spaces. Grayscale, hyperspectral, and medical pictures are the three kinds of images.

II. SEGMENTATION TECHNIQUE

Image segmentation may be divided into two categories:

- 1. Separation at the local level
- 2. Worldwide division

The term "worldwide division" refers to the process of segmenting a whole picture. Global segmentation is primarily concerned with segments having a high number of pixels. As a result, estimated parameter values for global segments are the most reliable. Three distinct philosophical viewpoints may be used to approach image segmentation[15].

If the pixel is part of an object, it has a value of one; otherwise, it has a value of zero. Segmentation is the process of separating low-level image processing from picture analysis. The pixel belongs to the object once the segmentation process is completed.

- 1. Hybrid Techniques
- 2. Structural Techniques
- 3. Stochastics Techniques

Structural techniques utilize information about the region's structure to segment it. Stochastic methods are used on discrete pixels without understanding or taking into account the region's structural information. The stochastic technique is based on a number of methods, one of which being statistical analysis. Hybrid methods are those that combine the best features Hierarchical and random approaches are also used[16].

A. Separation based on border recognition:

The strength information in a picture only gives you a sliver of information about where the edges are. The edge detection method involves locating a pixel on the area's edge. This technique tries to resolve picture segmentation by identifying edges or pixels amongst distinct areas with fast intensity transitions, which are then removed and connected to create closed object borders[17]. A binary image is the end outcome. The presence of noise generated during the imaging process, as well as afterwards in the broadcast and specimen procedures, is one cause of uncertainty. The fact that every measuring instrument is imperfect and its findings represent just partial observation is another source of uncertainty. This implies that edge detection techniques are often ill-posed, or restricted, and therefore may not have unique solutions. Looking for areas in an image where the intensity varies quickly, using one of these criteria, is the simplest method to identify edges in an image. The edge detection method is a structural tool used in picture segmentation. The gray histogram and the gradient based technique are the two major edge based segmentation methods, according to theory. The edges are discovered first, then connected together to create the necessary boundaries in the edge method[18]. Edge detectors use a variety of operators to detect edges, including the Others include the Sobel operators, the Laplace technician, the Clever driver, and the LoG operator. A higher image quality is required for the edge detection approach, thus noise must be reduced or removed[19].

The pixel values are one of the most basic approaches of segmentation. The method employs thresholding-based segmentation, which may aid in basic region-growing stages. Thresholding methods may be chosen depending on a priori information, whether explicitly or mechanically Threshold techniques that are based on edges, regions, or bothare the three types of threshold algorithms. The edge information is linked to edge-based algorithms. Edge points may be used to represent an object's structures. This kind of area includes edge detection techniques like the The Interpolation edge sensor and the canny edge detector are two types of edge detectors. These techniques are used to find edge pixels while reducing the noisy impact. Thresholding is an ancient, basic, and widely used picture segmentation method. For segmenting pictures with bright objects on a dark backdrop, image segmentation by thresholding is a basic but effective method. The method of thresholding is based on image space regions, or picture properties. Thresholding converts a multilayer picture to a It chooses an appropriate threshold T to partition image pixels into numerous sections and differentiate objects from background in a binary picture. Thresholding is a mechanism for dividing desire classes by defining an intensity value called a threshold. All photons with a brightness greater than the cutoff are assigned to one category, while the rest are assigned to another [20]-[22]. There are two kinds of thresholding techniques for selecting a thresholding value: global and local thresholding. Technique for image thresholding. Bi-level thresholding and multi-level thresholding are two types of thresholding. The method is known as global thresholding when T is constant, while it is known as local thresholding when T is variable. When the backdrop illumination is uneven, global thresholding techniques may fail. Multiple thresholds are employed in local thresholding to adjust for uneven light. Threshold selection is usually done interactively, although automated threshold selection methods may be derived. The thresholding method's limitations include the generation of just two classes, inability to apply to multichannel pictures, and sensitivity to noise and intensity in homogeneities.

B. Region Based Segmentation Methods

A connected homogeneous subset of an image indicated by R is defined as a connected homogeneous subset of the picture in terms of some criteria such as gray level or texture. In a picture, a region is a collection of linked pixels having comparable characteristics. Each pixel in the region approach is allocated to a certain item or area. Geographical area segmented methods are cheaper and less speckle than edge detection techniques. Edge-based approaches split an image into comparable areas based on a set of predefined parameters, while region-based approaches split a picture into comparable areas depending on a set of given requirements. Pixels that belong to an item are clustered together and indicated in region-based segmentation. The adoption of suitable thresholding methods is also required for region-based segmentation. and geographic closeness are key Value similarity concepts [23], [24]. The following techniques are used in region-based segmentation algorithms:

a. Region Growing

This criteria is based on strength data. The region extending technique of segmentation analyses neighboring images and assigns these to a zone class with no defined bounds. For each square in the town's boundary, this method is repeated. If nearby regions are found, a geographical area technique is used to melt edges whilst maintaining strong edges.

The method to expanding in a region is straightforward. The region boundaries discovered via region growth are precisely thin and linked. In terms of noise, the algorithm is likewise extremely steady. The only drawback is that it requires a seed point, which usually necessitates human intervention. As a result, a seed point is required for each divided area.

b. Separation and merger of regions:

Split and merge is the polar opposite of region growth. This method affects the whole picture. A top-down method to region splitting is used. It starts with a single picture and splits it into sections that are more homogeneous than the entire. Splitting is inadequate for good segmentation since it significantly restricts segment forms. As a result, a merging step following splitting is always desired, resulting in the split and-merge method. Any area may be divided into subregions, which can then be combined into a single region. Rather of selecting seed points, the user may split a picture into a series of disconnected areas and then combine them to try to meet the criteria of acceptable image segmentation. Theory based on quad tree data is often used to divide and merge regions[25].

C. Clustering-based classification:

Clustering, also known as data grouping, is an important first step in image processing. Clustering is an unsupervised learning problem in which pixels are classified by identifying a limited number of categories known as clusters. Clustering does not require any training steps and instead trains itself using existing data. When classes are known ahead of time, clustering is utilized. Similar pixels are clustered together to create clusters once a similarity criterion is established between them. The maximization of intra-class similar and maximization of inter-class similarities concepts is used to arrange pixels into clusters. Clustering is a method for determining the connection between patterns in a set by grouping them into groups or clusters, with patterns within a cluster being more similar to each other than patterns from other clusters. The product's measure, as well as its implementation, impact the quality of a clustering. Inside the identical group, an effective clustering algorithm will produce high groups with high intra-class resemblance and low interclass resemblance. Dissimilarity to other clusters' items. The method's similarity measure and its implementation both influence the quality of clustering results. The capacity to discover is also a criterion for a clustering method's quality. Clustering is the categorization of things into groups based on specific characteristics. In clustering methods, an effort is made to extract a vector from the image's local regions. Assigning each pixel to the closest cluster mean is a typical clustering technique. Hard clustering, k-means clustering, fuzzy clustering, and more types of clustering methods exist. Hard clustering, k-means clustering, fuzzy clustering, and other clustering techniques are categorized. The k-means method is a well-known and often used hard clustering technique (noted HCM). In hard cluster, every motif data is assigned a participation value of zero or one. Its operation is straightforward: it generates an initial hard cpartition, computes the c center, and assigns each item to the center closest to it in order to reduce within-cluster variation. It does a test after each iteration, comparing the current and previous partitions, and if the difference is less than a certain threshold, it quits; otherwise, it continues. The statistical clustering method k-means is used.

III. DISCUSSION

A neuronal net is a computerized simulation of the biological brain that tries to replicate the learning process. A neuronal networks is a kind of network, sometimes known as a neural net, is a kind of artificial neural network. Artificial neural networks (ANNs) have been extensively utilized to address the issue of medical picture segmentation in recent years. A vast number of parallel nodes make up a neural network based on life modeling, particularly the learning process of the human brain. Each node is capable of doing some fundamental computations. The learning process is accomplished by transferring connections and connection weights across nodes. Its primary benefit is that it is not reliant on the probability density distribution function. When the data deviates from the norm, it may also be used to prove the segmentation findings. During the picture segmentation process, a neural network may also minimize the need for expert involvement. Many age segmentation techniques suffer from this issue. To begin with, the picture segmentation problem is transformed into problems of energy reduction or classification, and so on. The problems are then addressed using a neural network in this manner. To identify the connectivity and weights between the nodes, the neural network was trained using a training sample set. The fresh pictures were then segmented using a neural network that had been trained. The feature extraction and picture segmentation using a neural network are two essential stages in the neural network segmentation technique.

IV. CONCLUSION

This survey of image segmented research gives a quick summary of the many segmentation approaches utilized in digital picture analysis. The paper also examines the research on different image segmentation research methods as well as other research problems in this area. These techniques are crucial for pattern identification and recognition utilizing edges, pictures, and points. Many sophisticated machines utilize the image segmentation methods discussed in this review article to identify faces, pictures, and recognize patterns. In medical research, image segmentation is used to identify malignant cells in medical pictures. They can also recognize roadways via satellite photos. As the universal segmentation algorithm, image segmentation has a bright and difficult future ahead of it, and it has been the focus of recent research. There is no one technique that is suitable for all types of pictures, nor are all methods suitable for all types of images. Due to the abovementioned causes, picture segment remained a challenging problem in image analysis and machine vision, and it is still a looming challenge throughout the world. Still image segmentation offers extra tools that might be utilised to a number of situations.

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