



A Survey on Advanced Image Segmentation Techniques

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How to cite this paper:

Astha Srivastava¹, Ankita Singh², Ankita Gupta³,
Rajkishore Yadav⁴, "A Survey on Advanced
Image Segmentation Techniques"
IJIRE-V3I02-294-297.

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Abstract: The Image segmentation is termed as an important step in image processing as a major step in this process. Image Segmentation is a way of clarifying the graphical/pictorial details and turn them into individual that is more consequential and obvious to examine by partitioning the image into numerous chunks (sets of pixel, also known as image entity), according to their features and characteristics. Sectionalization is used to separate the parts from image for further processing. Segmentation is main principle for the successful extraction of image property and their following categorization. Image processing approaches can be grouped into six classes: amplitude thresholding, component labelling, boundary-based segmentation, region-based segmentation, cluster analysis, clustering based segmentation, convolution, edge based, template matching, and surface segmentation. Impressive segmentation techniques are presently available; however, each technique is hasty. In this report, we are going to study about new and advanced modern methods for image classification in digital vision system and its future scope. Segmentation is a key for understanding the images.

Keywords: Image segmentation, Convolutional Neural Network, Rcn

I. INTRODUCTION

The Computer perception is a swiftly broadening area that is dependent on the ability to customarily segment, identify, elucidate, and explain fragmented images. During segmentation, an image is pre-processed which includes re-creation, improvement, amplification, portrayal or simple depiction of data. Certain attributes are drawn out; to partition the image into its chief constituent. The whole image is straight away fragmented into the simpler entity on the similarity/dissimilarity of edges, threshold value, color, texture, patterns, shapes, etc. The segmented depiction is dispelled to a classifier or to an image-modelling system. The image classification operation maps different regions or portions into one or several objects. Each object is identified by a stamp/label. The image-understanding structure then decides the connection between different objects in a scene to provide a complete scene illustration. This Paper covers the most recent literature in image segmentation.

Introducing the associated research paper on image segmentation by various researchers and authors with distinct viewpoint with regards to the segmentation process:

Satish Kumar, in his study described about the various applications that use the theory of the image segmentation that involves, computer science, machine vision, surveillance, medical, scanning, recognition, detection etc.

Rajeshwar Dass et al., In his survey classified and explained main image segmentation algorithms and concluded that the process are assigned on the basis of the characteristics as: homogenous nature, spatial traits of the continuity in image, texture, content. Nikita Sharma in their paper, evaluated that the performance judgement and differentiation are not easy for the several segmentation methodology or manners. They mentioned the evaluated fashion so that researchers could use it while selecting any segmentation techniques.

P. Sravani, in their survey, gave an outline of various segmentation methods and clustering are studied. Although many approaches have been developed, not all of them are useful for all types of images. Segmentation segments the image and clusters them together according to some similarity. Distance metric is one that measures the similarity and has direct impact on the formed cluster.

Fuzzy is a powerful unsupervised clustering technique which is widely used for robust segmentation of real time images. Traditionally FCM and many other algorithms use Euclidean Distance metric. H. P. Narkhede in his research of image segmentation study, described various methodologies and issues concerning the digital process used in many recognition patterns.

Punam Thakare, described different image segmentation techniques and discusses in detail the edge detection techniques and their evaluation. It gives an algorithm which is a combination of detection and evaluation of the edge detectors. The results show that the recognition rate depends on the type of the image and their ground truth.

II. ADVANCED DEEP NEURAL NETWORK TECHNIQUES.

Deep neural networks (DNN) is an improved versions of the conventional ANN with multiple layers. With the increasing demand of deep learning over the past few years, several other neural architectures have been proposed, such as feed – forward neural network, recurrent neural network, gated recurrent units, capsule networks, , spatial transformer networks, single node with its own feedback etc. The DNN models are recently becoming very popular due to its excellent performance to learn the nonlinear input–output mapping also including the underlying structure of the input data vectors . DNN gives model an scope for improvement in its effectiveness and accurate results. This permits a model to take a set of inputs and give an output. Using Neural Network is a simple methods for image processing similar to as copying and pasting the codes. It doesn't depend on the what ML platform we use; directing it to use multiple nodes at each layer is the main task, and quite easy to perform. The Deep Network lets the model to do inference the image information in the hidden layer, called the black box. The black box is hard to investigate.

III. CONVOLUTIONAL NEURAL NETWORK

A convolutional neural network (CNN) is a type of deep neural networks(DNN), commonly used to verify visualization. Now when we think of a neural network we think about matrix multiplications but that is not the case with Conv Net. It uses a special technique called Convolution CNNs which are among the most productive and widely acceptable design in the deep learning for image processing , especially for computer visualization.

CNNs were initially proposed by Fukushima in his seminal paper on the “Neocognitron” based on the hierarchical models of visual system proposed by Hubel and Wiesel. This techniques is neural network based with layered topology. Waibel also ; introduced CNNs with weights shared among temporal receptive fields and back propagation training for phoneme recognition, and LeCun et al. introduced a CNN framework for document identification .

CNNs mainly consist of three layers:

- i) convolutional layers – It is the core building block of cnn . It performs the dot product between two matrices, where first matrix is a set of learnable parameters known as kernels while other is restricted portion. Example : having an input of size $X \times X \times W$ and D out number of kernels with spatial size of F with stride S and padding P then

$$W_{out} = \frac{X - F + 2P}{S} + 1$$

- ii) Non-linearity layers: Convolution is a linear operation, but this non linearity is placed after convolution layer to represent the non – linearity of activation map.
- iii) pooling layers: This layer replaces the output at certain places by summarizing statistic to the nearby output. This helps in minimizing the spatial size of presentation, which further decreases the computation and weights required . By stacking layers to form multi-resolution pyramids, the higher-level layers learn features from

increasingly wider receptive fields. The main computational advantage of CNNs is that all the receptive fields in a layer share weights, resulting in a significantly smaller number of parameters than fully-connected neural networks.

Example: If we an activation map of size $X \times X \times W$, a pooling pf size F , and stride S then the output volume size is

$$W_{out} = \frac{W - F}{S} + 1$$

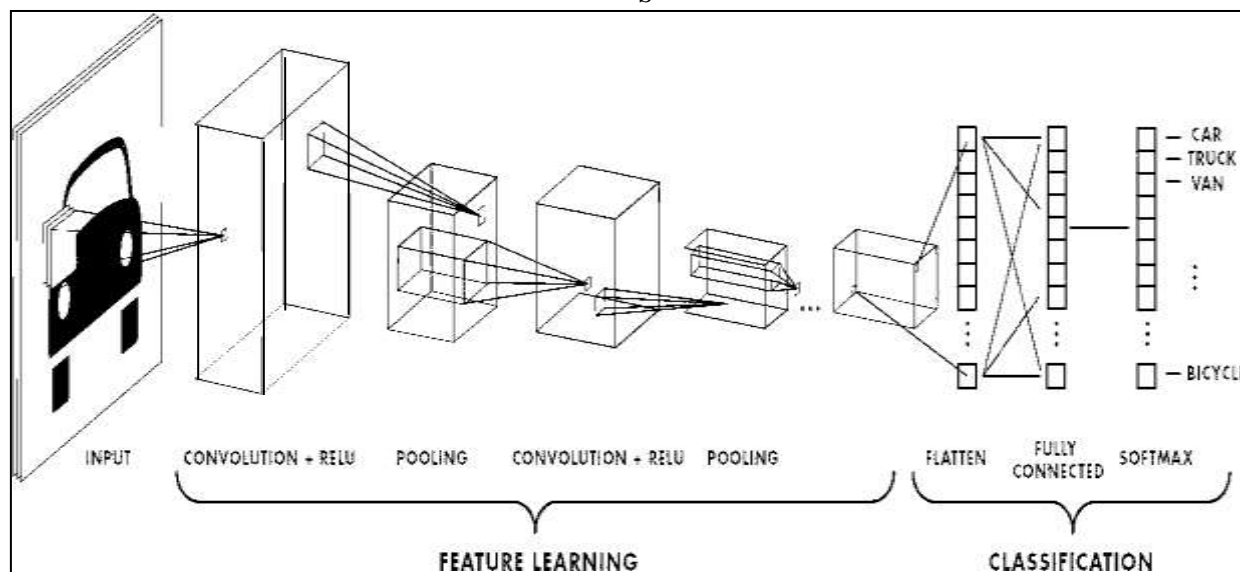


Fig 1: Process of Convolutional Neural Network

IV. REGIONAL CONVOLUTIONAL NEURAL NETWORK

RCNN stands for regional convolutional network and it is a concept based on region proposal. Its extensions like Fast R-CNN, Faster R-CNN, Mask-RCNN had shown great advancement in successful object detection applications. In particular, the Faster R-CNN architecture developed for object detection uses a region proposal network (RPN) to propose bounding box candidates. The RPN draws out the Region of Interest (RoI), and RoI Pool layer determine features from these proposals in order to conclude the bounding rectangle coordinates and the objective of the object. Some of the extensions of R-CNN have been labouriously used to examine the instance segmentation problem; i.e., the task of concurrently performing object detection as well as semantic segmentation.

In one of the extension of this model, He .. proposed a Mask R-CNN for image instance segmentation, which strike all the previous standard on many COCO challenges. This technique efficiently detects objects in an image while at the same time generating a high-quality decomposition mask for each instance. Mask R-CNN is essentially a Faster RCNN with 3 output branches -the first computes the bounding rectangle coordinates, second branch figure out the associated classes, and the third determine the binary mask to segment the object. The Mask R-CNN loss function combines the losses of the bounding box coordinates, then predict lass, and the segmentation mask, and trains all of them jointly.

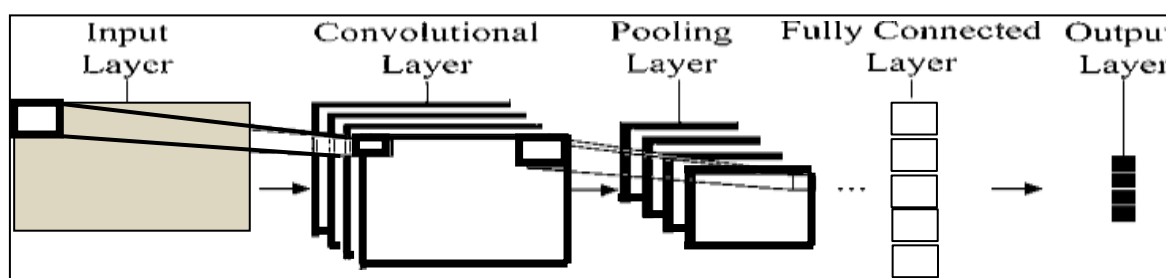


Fig 2: R-cnn layer for Visualization.

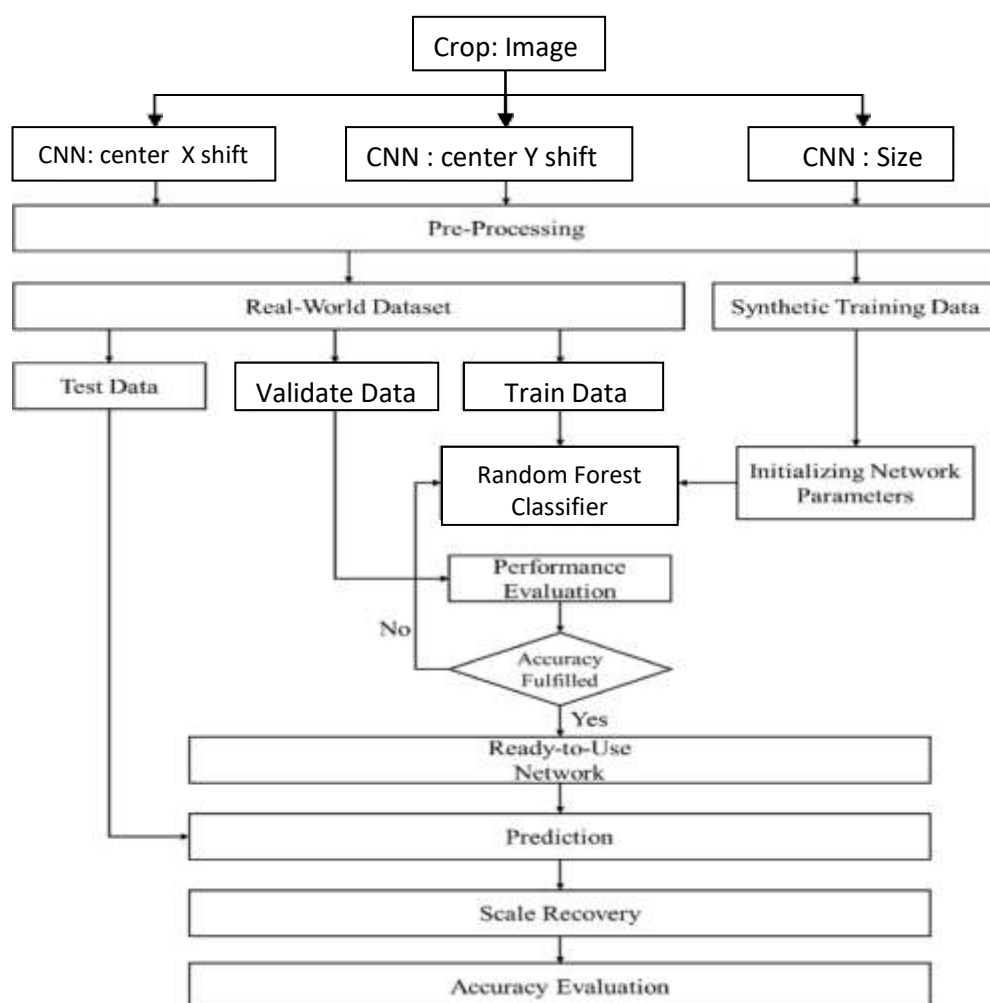


Fig 3: Flowchart of Convolutional Neural Network training Data.

II. CONCLUSION

Concluding the paper, we studied about the latest image segmentation i.e based on neural network known as convolution neural network. DL has provided as methods to compute the once's difficulties in simultaneously/concurrent modeling multi-complex modules of raw data. In recent years Deep Learning developments, another common approach is that of multi-modal DL. We conclude that CNNs can be used to solve both estimation and categorization of the visual scene.

- DL requires sizeable datasets or trained data to analyse unspecified data and to train the operation for accuracy. This turns into a challenge sometimes when it is particularly difficult when real-time data processing is required or when the received datasets are not brief and limited.
- The utilization of in-depth and various structural adaptations is crucially improved in the CNN learning capacity. Switching the customary layer configuration with block results advances CNN performance, presently, developing novel and efficient architectures is the main concept in new research model's of CNN architectures.
- It is expected that cloud-based applications will play an important role in the future progress of computational DL applications. Utilizing cloud computing benefits a solution by handling the large amount of data.

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