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# A Comprehensive Review of Modern Object Segmentation Approaches

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# Foundations and Trends<sup>®</sup> in Computer Graphics and Vision

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# A Comprehensive Review of Modern Object Segmentation Approaches

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## ABSTRACT

Image segmentation is the task of associating pixels in an image with their respective object class labels. It has a wide range of applications in many industries including healthcare, transportation, robotics, fashion, home improvement, and tourism. Many deep learning-based approaches have been developed for image-level object recognition and pixel-level scene understanding — with the latter requiring a much denser annotation of scenes with a large set of objects. Extensions of image segmentation tasks include 3D and video segmentation, where units of voxels, point clouds, and video frames are classified into different objects. We use “Object Segmentation” to refer to the union of these segmentation tasks. In this monograph, we investigate both traditional and modern object segmentation approaches, comparing their strengths, weaknesses, and utilities. We examine in detail the wide range of deep learning-based segmentation techniques developed in recent years, provide a review of

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the widely used datasets and evaluation metrics, and discuss potential future research directions.

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# 1

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## Introduction

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### 1.1 Overview

Automated visual recognition tasks such as image classification, image captioning, object detection and image segmentation are essential for image and video processing. Before the advent of deep neural networks, traditional techniques leveraged hand-crafted heuristics to extract visual features and manually tuned parameters to combine these features for inferences and decisions. These techniques are simple yet effective for many cases. However, they are often not generalizable and difficult to configure. For example, Canny Edge Detection, an algorithm that uses Gaussian filters and thresholding to identify edges in images have two key adjustable parameters-size of filters and edge strength thresholds that need to be tuned. If not selected carefully, these parameters can greatly impact the effectiveness of the algorithm, resulting in either missing or false positive edges.

### 1.2 Convolutional Neural Networks

With the advent of the deep learning era, more powerful and accurate automated visual recognition techniques were enabled by deep neural

networks that can handle high-dimensional data. In particular, convolutional neural networks (CNN) (LeCun *et al.*, 1999) excel at image recognition and pattern identification due to its ability to capture space invariance of shapes in images. CNNs achieve this through two major components: the convolutional layer, pooling layer, and non-linear layers. The convolutional layers apply sliding kernels to extract features from the input image, starting from low-level features such as edges in the earlier layers, to high-level features such as eyes and nose in the later layers. The pooling layer then reduces the resolution of the output feature maps from the convolutional layers, allowing the network to achieve translational and deformation invariance. LeNet-5, developed by LeCun *et al.* (1998), is the first convolutional neural network applied to the famous handwritten digit recognition task MNIST.<sup>1</sup> Although only consisting of a pair of consecutive convolutional and pooling layers followed by three fully connected layers, LeNet-5 is able to achieve a 98.49% accuracy on the MNIST dataset. However, when applied to The ImageNet Large Scale Visual Recognition Challenge (ILSVRC) (Deng *et al.*, 2009) task, the test accuracy drops to only 66%. Compared with the human level accuracy of 94%, LeNet-5 is not sufficient for more challenging visual recognition tasks.

In 2012, a breakthrough was made when Hinton's team won the ILSVRC challenge with deep learning (Krizhevsky *et al.*, 2012). The challenge included the largest image dataset of the time, totaling over one million images spanning 1,000 object categories. For the first three years since the challenge commenced, most visual recognition systems did not make any breakthrough at this image classification task. The deep learning model proposed by Krizhevsky *et al.* (2012) was named AlexNet, which is often considered the first deep convolutional network. It is similar in structure with LeNet-5, but consists of 3 more convolutional layers and has a total of 62 million trainable variables. It significantly boosted the performance of previous state-of-art image classification techniques by achieving considerably better classification accuracy in the ILSVRC challenge than the second place, reducing the top-5 error rate by over 10% (Krizhevsky *et al.*, 2012). It was not until then that

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<sup>1</sup><http://yann.lecun.com/exdb/mnist/>

deep learning started to take over as the go-to approach for challenging computer vision tasks.

Despite the significant improvement AlexNet made towards image classification, it uses too many parameters and is therefore intractable for large-scale training. VGGNet (Simonyan and Zisserman, 2014) solves this problem by replacing the large kernel-sized convolutional layers with multiple layers containing  $3 \times 3$  kernels. The network contains a total of 13 convolutional layers and 3 fully connected layers. The authors demonstrate that many small-sized filters achieve the same functions as fewer filters with large kernel sizes, but with much fewer parameters.

As CNNs continue to develop and grow deeper, a problem that often arises during the training stage is vanishing gradient – weights disappearing during back-propagation, which hinders the performance of models. ResNet (He *et al.*, 2016), developed one year after VGGNet (Simonyan and Zisserman, 2014), leverages two types of skip connections – identity and projection, to tackle this issue. In addition, it also uses batch normalization (Ioffe and Szegedy, 2015) to stabilize and increase training speed.

Developed by Google, The Inception (Szegedy *et al.*, 2015) network is aimed to capture sparse salient features of varying sizes. It achieves this through stacking multiple sizes ( $5 \times 5$ ,  $3 \times 3$ , and  $1 \times 1$ ) of kernels at the same layer and combining outputs to feed to the next layer.  $1 \times 1$  convolutions are used to reduce the number of channels, thereby reducing the number of parameters of the network.

### 1.3 Object Detection

A precursor to the image segmentation problem, object detection not only classifies, but also localizes each object in an image. One of the earlier approaches to solving object detection is the sliding window approach (Bosch *et al.*, 2007; Chum and Zisserman, 2007; Dalal and Triggs, 2005; Ferrari *et al.*, 2007; Rowley *et al.*, 1995). By taking different crops from an input image and applying CNNs to classify cropped regions (a cropped region is classified as “background” if no target object is identified), different objects are simultaneously localized and categorized from the input image.

However, because objects appear in various sizes and in many locations in the image, it is very computationally expensive to apply CNNs to a potentially infinite set of cropped input regions. Instead, a more scalable approach called region proposals (Alexe *et al.*, 2012; Uijlings *et al.*, 2013; Zitnick and Dollár, 2014) was developed to find candidate regions in images that are likely to contain objects quickly based on traditional image and signal processing techniques. One example of such a technique is Selective Search, which uses sub-segmentation and greedy search to generate and recursively combine regions into a final list of 2,000 candidate regions (Uijlings *et al.*, 2013). With region proposals approaches, CNNs can then be applied to a much smaller set of cropped regions to produce final detection outputs, thus making this task much more computationally tractable.

The idea of combining region proposals and CNNs for object detection first materialized in Girshick *et al.* (2014). After the features are extracted from the proposed regions using CNNs, an SVM is used to 1) determine whether an object is present in the proposed region and classifies it if present, and 2) perform bounding box regression to compute offsets to the proposed region and refine its boundary. Although R-CNN is able to resolve the challenges with the sliding window approach to object detection, it is still very slow since CNNs need to be applied to 2,000 regions, and selective search could generate bad proposals since it is a fixed algorithm.

The same authors of R-CNN developed Fast R-CNN (Girshick, 2015) to address the slow selective search stage. Instead of applying CNNs to proposed regions, the input image is fed into the CNNs directly to generate a feature map. From the feature map, candidate regions are produced and an RoI pooling layer is then applied to reshape these regions into fixed sizes before they are fed into a fully connected region to produce a feature vector. Finally, a softmax layer is applied to classify this feature vector into objects, and a bounding box regressor is used to compute the offset values to refine boundaries. This is a much faster approach than R-CNN because the CNNs need only be run one time on the input image instead of on 2,000 regions. Although Fast-R-CNN is significantly faster than R-CNN during both training and testing, the region proposal stage remains to be the bottleneck. This is because it

still uses selective search on the CNN feature map to propose candidate regions.

Faster R-CNN (Ren *et al.*, 2015) is a similar approach to both algorithms above, but leverages a separate network on the CNN feature map to propose candidate regions instead of using selective search. As a result, Faster R-CNN beats its predecessors in inference speed and can generate detection results in real-time.

Whereas the object detection models from the R-CNN family use regions to identify objects in an image, YOLO (Redmon *et al.*, 2016) is an approach that looks at not just the regions of interest, but the entire image. The image is divided into grids. A set of base boxes centered at each grid cell is fed as input to a single CNN to predict both object class scores and bounding boxes. This approach is extremely fast as detection is reframed as a regression problem. YOLO also sees the entire image when making predictions and makes fewer mistakes compared to Fast R-CNN when identifying background patches (Redmon *et al.*, 2016). However, it is not good at predicting small objects due to the constraints placed on the base boxes.

Similar to YOLO, SSD (Liu *et al.*, 2016a) is a single-shot (achieves object localization and detection simultaneously) object detector with fast inference speed to enable real-time detection. It uses a modified version of MultiBox (Erhan *et al.*, 2014), a bounding box regression technique, to generate candidate bounding boxes on each grid cell of the feature map. Given a set of default bounding boxes with different aspect ratios, SSD selects those with an intersection over union (IOU) score larger than 0.5 with the ground truth bounding box as candidates and performs bounding box regression to refine the boundaries. It also predicts, for each object class  $c$  and each candidate bounding box  $b$ , the probabilities of box  $b$  containing object  $c$ .

## 1.4 Object Segmentation

While object detection aims only to localize objects with bounding boxes and classifying localized objects, object segmentation adds the task of delineating precise object boundaries by classifying every unit in an input to an object class or the background. The general formulation for this



problem can be defined as follows: given an input space  $\mathcal{X}$ , and a label space  $\mathcal{L}$ , object segmentation finds a mapping  $f: \mathcal{X} \mapsto \mathcal{L}$ . It has a wide range of applications in many industries including healthcare (e.g. cancer detection), transportation (self-driving), robotics (guidance), fashion (virtual try-ons), home improvement (decor visualizers), and tourism (destination recognition). Segmentation of an image breaks it down into simplified representations that can aid downstream interpretation tasks. Thus, segmentation differs from detection as it does not have to be tied to a specific task, but rather an intermediate step in machine perception. Nevertheless, it is more challenging and usually more time-consuming than object detection, thereby requiring more advanced techniques and more high-quality annotated training data. Object segmentation is fundamentally tied to human image perception as recognition of objects from an input image can be conceptualized into a process of dissecting regions of deep concavity into simple volumetric components such as blocks, wedges, and cones (Biederman, 1985).

Recognition-by-components (RBC) theory states that robust object perception is possible because detection of edge properties such as curvature, symmetry, and parallelism is usually invariant over image quality and viewing angle (Biederman, 1985). However, because machines process images through various means unlike the process of human interpretation, it is far from enough to only focus on these components when performing object segmentation. In this monograph, we review various techniques developed over the years for the task of object segmentation, with a focus on the deep learning-based techniques for the two most widely solved segmentation tasks: Semantic Segmentation and Instance Segmentation. We also survey methods developed for Panoptic, Video and 3D data segmentation. We categorize and compare various techniques developed for each task. The readers will find that various themes emerge from these techniques that push machines to their limits and often times deviate from human perception principles. In addition, we provide an overview of the widely used benchmark datasets for each of these techniques, along with the respective evaluation metrics to measure the models' performances. Finally, we discuss potential directions for future research in these areas.

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