

UNIVERSITY OF MÜNSTER
DEPARTMENT OF INFORMATION SYSTEMS

Title

SEMINAR THESIS

in the context of the seminar

MORE THAN MEETS THE A-EYE: REFLECTING HUMAN VISION IN ARTIFICIAL
INTELLIGENCE

submitted by

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1 Introduction

1.1 Motivation and Context

1.2 Research Gap

2 Theoretical Background

2.1 Image Segmentation

Image segmentation is the process of dividing an image into different regions by grouping pixels and assigning each pixel a label. This step is an important part of many computer vision applications, such as detecting tumors in medical images or identifying pedestrians in autonomous driving. According to human visual perception, the identified regions are non-overlapping and meaningful - however, defining what exactly counts as a “meaningful” region can be difficult, as human perception is subjective and object boundaries are not always clear (Yu et al., 2023).

There are three common types of segmentation:

Semantic segmentation assigns every pixel in an image a semantic label, such as “car” or “sky”. *Instance segmentation* separates individual objects within the same class, for example distinguishing several people in one image. *Panoptic segmentation* combines both approaches by providing pixel-wise class labels and also identifying individual object instances.

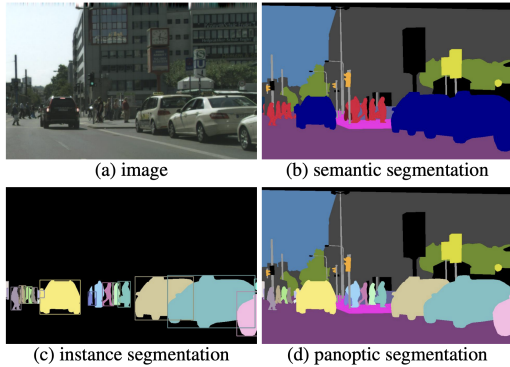


Figure 1 Types of image segmentation by Kirillov et al., 2019

Earlier approaches to image segmentation include algorithms such as k-means-clustering (Dhanachandra et al., 2015). Yet in recent years, deep learning models have significantly improved the segmentation effect and performance, therefore becoming the dominant method for solving segmentation tasks in complex environments (Minaee et al., 2022).

According to Zhou et al., 2024, the above-described image segmentation methods fall into the category of generic image segmentation (GIS). The category of promptable image segmentation (PIS) extends GIS by specifying the target to segment through a prompt. This prompt can have various forms such as text, box or point.

2.2 Salient Object Detection

The human visual system pays more attention to certain parts in an image, a property known as saliency. Inspired by this mechanism, *saliency detection* models aim to predict which regions in an image are most likely to attract human visual attention. These models typically provide saliency maps in form of heat maps, in which higher intensity values indicate regions detected to be more important (Ahmadi et al., 2018).

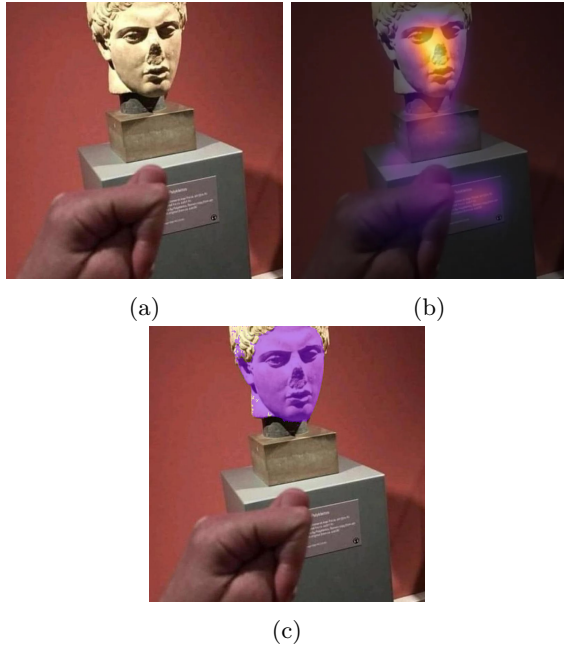
Salient Object Detection (SOD) – also referred to as salient object segmentation (Borji et al., 2019) or saliency segmentation (Kakanopas & Worarapanya, 2021) – goes one step further by segmenting the most salient object(s) of an image. SOD can be interpreted as a two-stage process: 1) Detection of the most salient object and 2) Accurate segmentation of the region of that object. In contrast to general image segmentation, SOD focuses on segmenting only those objects that are (or that are predicted to be) most salient (Borji et al., 2019; Liu et al., 2011). Figure 2 illustrates the difference between saliency detection and salient object detection.

2.3 Image Segmentation Models

TODO: noch überarbeiten, hier nur erste Ideen:

Figure 3 provides an overview of prominent image segmentation models.

Only promptable segmentation models are suitable for our use case, as our pipeline requires point-based segmentation based on eye-tracking



+ SAM3 release: Nov 20, 2025; in 2 weeks already 5.1k stars on github (more than SEEM in 2 years) - indicates strong community interest

+ SAM2: 52.8k stars on github

Figure 2 (a) the original image, (b) saliency map (Alexander Kroner, 2025) (c) salient object detection mask generated using SAM3 guided by the eye-tracking data

Generic Image Segmentation			Promptable Image Segmentation
Instance segmentation	Semantic segmentation	Panoptic segmentation	
Mask R-CNN	DeepLabV3	Mask2Former	SAM, SAM2, SAM3
YOLOv11-seg	FCN	Panoptic-DeepLab	Grounded-SAM
YOLACT	U-Net	Panoptic FPN	SEEM
Mask2Former	SegFormer	...	Florence-2
Mask DINO	Mask2Former		...
...	LR-ASPP		
	...		

Figure 3 Overview of prominent image segmentation models

data. Classic semantic or instance segmentation models do not support point-based object selection. Among promptable models, SAM (and its successors SAM2 and SAM3) is the only widely used model that natively supports point prompts. Grounded-SAM and Florence-2 offer text-promptable capabilities but do not add any advantages for point-based segmentation. SEEM provides multimodal prompting but is less actively maintained compared to SAM2/SAM3. Therefore, SAM is the most suitable model for our use case.

+ Cited over 15.000 times, making it the de facto standard for domain specific implementations

3 Methodology

3.1 Experimental Design

3.2 Planned Practical Steps

4 Results

4.1 Model Selection

4.2 Expected Results

A Appendix

TODO: Add result pictures and/or our code here

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