

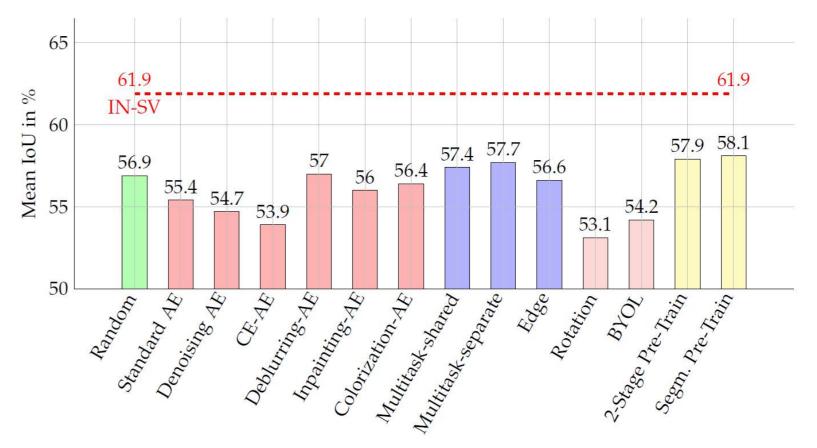
# **Self-Supervised Deep Representation Learning for Semantic Segmentation**

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## **Introduction and Motivation**

- Pre-training commonly used technique in computer vision; only done with ImageNet
- Costly and time-intensive annotation for semantic segmentation limits dataset size and applicability
- Low utilization of large amounts of relatively cheap unlabeled image data
- Standard pre-training has to overcome a task and domain gap harming the efficiency of pre-training

### **Results I**



## Approach

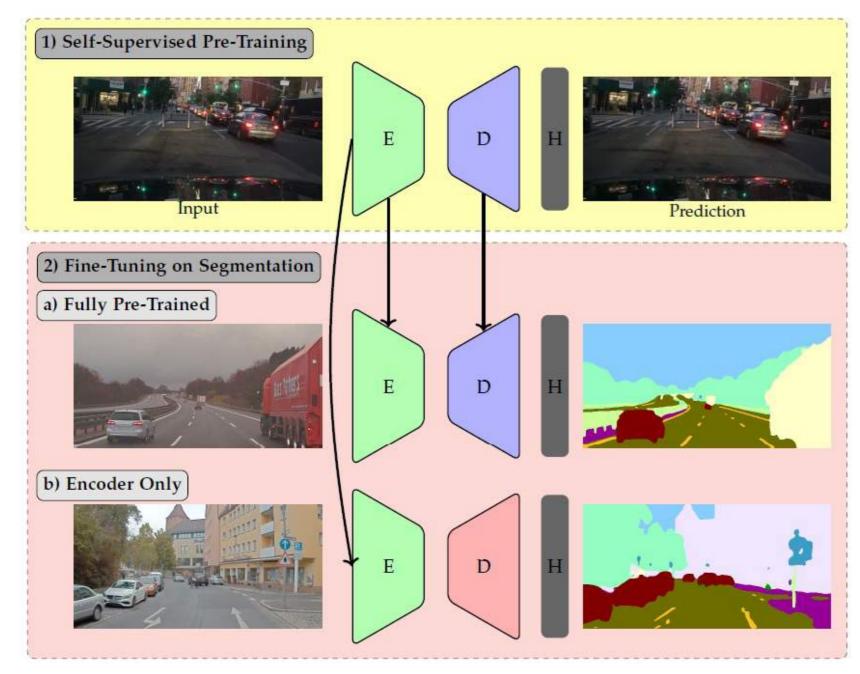


Figure 1: Self-Supervised Learning Approach

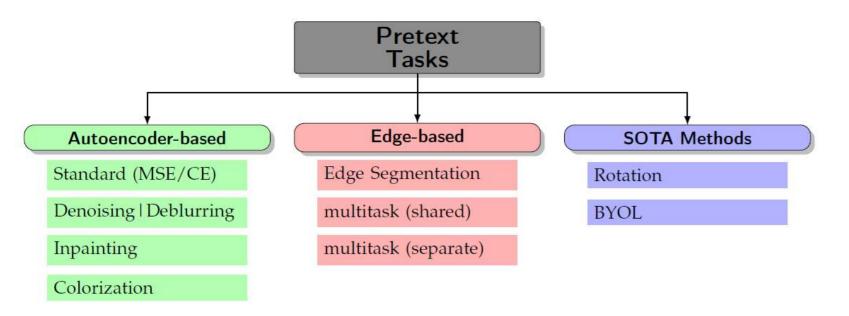


Figure 2: Pretext Tasks (© CARIAD SE)

Figure 4: mIoU for fine-tuning on Segmentation (© CARIAD SE)

- None of the evaluated pretext tasks performed competitive with ImageNet pretraining
- Only slight improvements over random initialization  $\rightarrow$  ineffecient pre-training

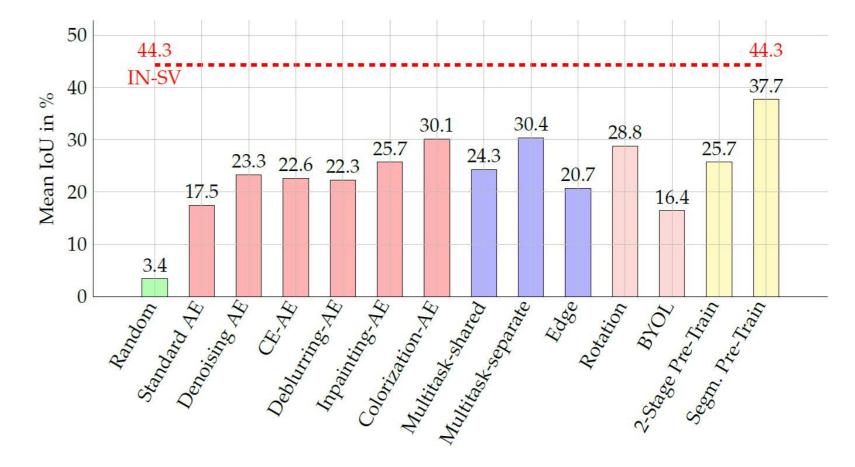


Figure 5: mIoU for frozen encoder (© CARIAD SE)

- Experiments with a frozen encoder revealed that the representations are semantically useful but not competitive with ImageNet pre-training
- Significantly better performance than random init; multitask performs best

Idea: self-supervised pre-training on automotive datasets like A2D2 or BDD100k

### **Pretext Tasks**

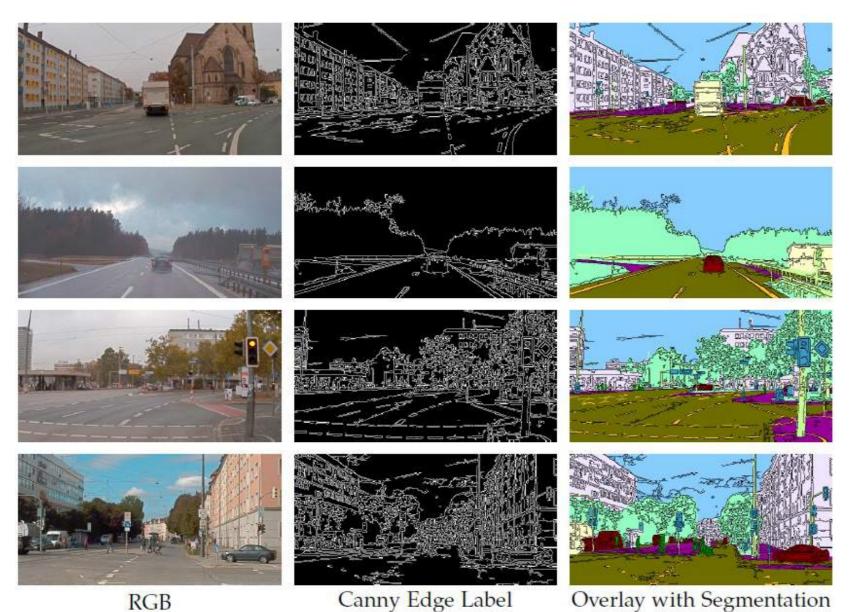


Figure 3: Edge Segmentation

- Semantic correspondence between canny edge detection and semantic segmentation
- Idea: Predict edges as pre-training  $\rightarrow$  useful bias for segmentation
- Also multitask pretext task (autoencoder + edge segmentation)

#### **Results II**

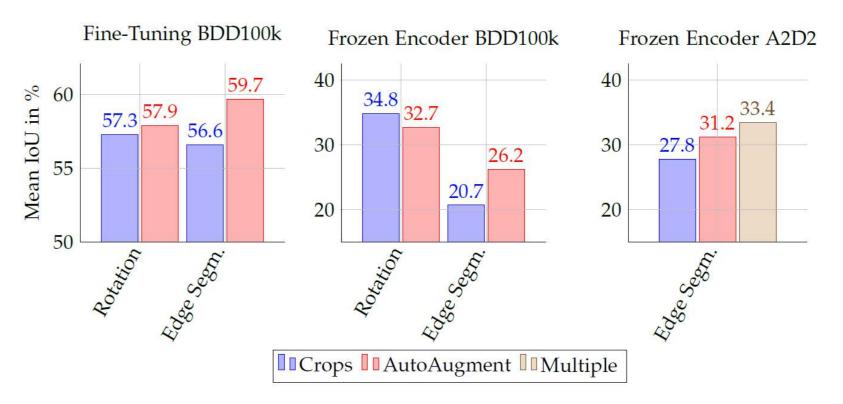
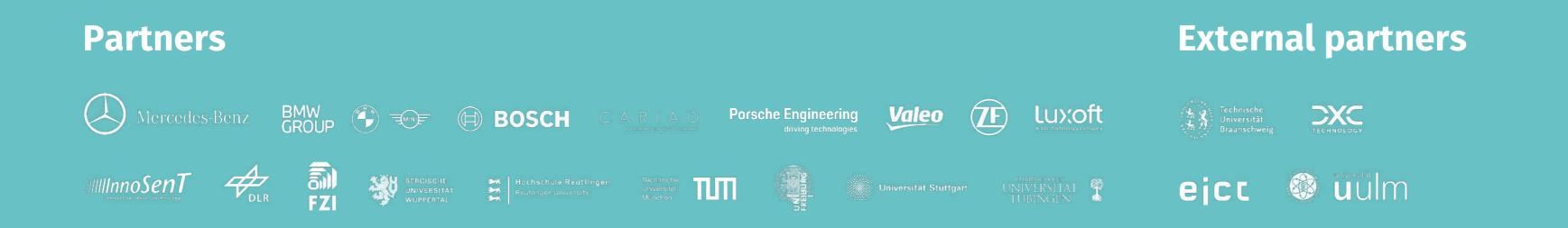


Figure 6: Pre-training under augmentations (@ CARIAD SE)

When using augmentations during pre-• training significant performance gains could be achieved across datasets and pretext tasks

## Conclusions

- Self-supervised pre-training on automotive datasets cannot compete with ImageNet pre-training
- Lots of different impacting factors
- Pre-training on one million images provides no gain; augmentations are promising



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