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Semantic Image Segmentation: Two Decades of Research

Gabriela Csurka

Naver Labs Europe

Gabriela.Csurka@naverlabs.com

Riccardo Volpi

Naver Labs Europe

Riccardo.Volpi@naverlabs.com

Boris Chidlovskii

Naver Labs Europe

Boris.Chidlovskii@naverlabs.com

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Semantic Image Segmentation: Two Decades of Research

Gabriela Csurka¹, Riccardo Volpi² and Boris Chidlovskii³

¹*Naver Labs Europe, France; Gabriela.Csurka@naverlabs.com*

²*Naver Labs Europe, France; Riccardo.Volpi@naverlabs.com*

³*Naver Labs Europe, France; Boris.Chidlovskii@naverlabs.com*

ABSTRACT

Semantic image segmentation (SiS) plays a fundamental role in a broad variety of computer vision applications, providing key information for the global understanding of an image. This survey is an effort to summarize two decades of research in the field of SiS, where we propose a literature review of solutions starting from early historical methods followed by an overview of more recent deep learning methods including the latest trend of using transformers. We complement the review by discussing particular cases of the weak supervision and side machine learning techniques that can be used to improve the semantic segmentation such as curriculum, incremental or self-supervised learning.

State-of-the-art SiS models rely on a large amount of annotated samples, which are more expensive to obtain than labels for tasks such as image classification. Since unlabeled data is instead significantly cheaper to obtain, it is not surprising that Unsupervised Domain Adaptation (UDA) reached a broad success within the semantic segmentation community. Therefore, a second core contribution of this monograph is to summarize five years of a rapidly growing

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field, Domain Adaptation for Semantic Image Segmentation (DASiS) which embraces the importance of semantic segmentation itself and a critical need of adapting segmentation models to new environments. In addition to providing a comprehensive survey on DASiS techniques, we unveil also newer trends such as multi-domain learning, domain generalization, domain incremental learning, test-time adaptation and source-free domain adaptation. Finally, we conclude this survey by describing datasets and benchmarks most widely used in SiS and DASiS and briefly discuss related tasks such as instance and panoptic image segmentation, as well as applications such as medical image segmentation.

We hope that this monograph will provide researchers across academia and industry with a comprehensive reference guide and will help them in fostering new research directions in the field.

Preface

Semantic image segmentation (SiS) plays a fundamental role towards a general understanding of the image content and context. In concrete terms, the goal is to label image pixels with the corresponding semantic classes and to provide boundaries of the class objects, easing the understanding of object appearances and the spatial relationships between them. Therefore, it represents an important task towards the design of artificial intelligent systems. Indeed, systems such as intelligent robots or autonomous cars should have the ability to coherently understand visual scenes, in order to perceive and reason about the environment holistically.

Hence, semantic scene understanding is a key element of advanced driving assistance systems (ADAS) and autonomous driving (AD) (Teichmann *et al.*, 2018; Hofmarcher *et al.*, 2019) as well as robot navigation (Zurbrügg *et al.*, 2022). The information derived from visual signals is generally combined with other sensors such as radar and/or LiDAR to increase the robustness of the artificial agent's perception of the world (Yurtsever *et al.*, 2020). Semantic segmentation fuels applications in the fields of robotic control and task learning (Fang *et al.*, 2018; Hong *et al.*, 2018b), medical image analysis (see Section 4.3), augmented reality (DeChicchis, 2020; Turkmen, 2019), satellite imaging (Ma *et al.*, 2019) and many others.

The growth of interest in these topics has also been caused by recent advances in deep learning, which allowed a significant performance boost in many computer vision tasks – including semantic image segmentation. Understanding a scene at the semantic level has long been a major topic in computer vision, but only recent progress in the field has allowed machine learning systems to be robust enough to be integrated into real-world applications.

The success of deep learning methods typically depends on the availability of large amounts of annotated training data, but manual annotation of images with pixel-wise semantic labels is an extremely tedious and time consuming process. As the major bottleneck in SiS is the high cost of manual annotation, many methods rely on graphics platforms and game engines to generate synthetic data and use them to train segmentation models. The main advantage of such synthetic rendering pipelines is that they can produce a virtually unlimited amount of labeled data. Due to the constantly increasing photo-realism of the rendered datasets, the models trained on them yield good performance when tested on real data. Furthermore, they allow to easily diversify data generation, simulating various environments and weather/seasonal conditions, making such data generation pipeline suitable to support the design and training of SiS models for the real world.

While modern SiS models trained on such simulated images can already perform relatively well on real images, their performance can be further improved by domain adaptation (DA) – and even with *unsupervised domain adaptation* (UDA) not requiring any target labels. This is due to the fact that DA allows to bridge the gap caused by the *domain shift* between the synthetic and real images. For the aforementioned reasons, sim-to-real adaptation represents one of the leading benchmarks to assess the effectiveness of *domain adaptation for semantic image segmentation* (DASiS).

The aim of our monograph is to overview the research field of SiS. On the one hand, we propose a literature review of semantic image segmentation solutions designed in the last two decades – including early historical methods and more recent deep learning ones, also covering the recent trend of using transformers with attention mechanism. On the other hand, we devote a large part of the monograph to survey methods

designed *ad hoc* for DASiS. While our work shares some similarities with some of the previous surveys on this topic, it covers a broader set of DASiS approaches and departs from these previous attempts pursuing different directions that are detailed below.

Amongst the existing works surveying SiS methods, we can mention Thoma (2016) who gives a brief overview of some of the early semantic segmentation and low-level segmentation methods. Li *et al.* (2018a) and Zhou *et al.* (2018) discuss some of the early deep learning-based solutions for SiS. A more complete survey on deep SiS models has been proposed by Minaee *et al.* (2020), while Zhang *et al.* (2020a) focus on reviewing semi- and weakly supervised semantic segmentation models. We cover most of these methods in Section 1, where we provide a larger spectrum of the traditional SiS methods in Section 1.1. Then, in Section 1.2, we organize the deep SiS methods according to their *most important characteristics*, such as the type of encoder/decoder, attention or pooling layers, solutions to reinforcing local and global consistency. In contrast to the previous surveys, this section also includes the latest SiS models that use attention mechanisms and transformers as encoder and/or decoder. One of the core contributions of this section is Table 2.1, which presents a broad set of deep models proposed in the literature, and summarized according to the above mentioned characteristics. Finally, in Section 1.3 we review not only semi- and weakly supervised SiS solutions, but also new trends whose goal is improving semantic segmentation, such as curriculum learning, incremental learning and self-supervised learning.

In Section 2, we present and categorize a large number of approaches devised to tackle the DASiS task. Note that previous DA surveys (Gopalan *et al.*, 2015; Csurka, 2017; Kouw and Loog, 2021; Zhang and Gao, 2019; Venkateswara and Panchanathan, 2020; Singh *et al.*, 2020; Csurka, 2020; Wang and Deng, 2018; Wilson and Cook, 2020) address generic domain adaptation approaches that mainly cover image classification and mention only a few adaptation methods for SiS. Similarly, in recent surveys on domain generalization (Wang *et al.*, 2020b; Zhou *et al.*, 2020a), online learning (Hoi *et al.*, 2018) and robot perception (Garg *et al.*, 2020), several DA solutions are mentioned, but yet DASiS received only marginal attention here. The most complete survey – and therefore most similar to the content of our Section 2 –

is by Toldo *et al.* (2020a), which also aimed at reviewing the recent trends and advances developed for DASiS. Nevertheless, we argue that our survey extends and enriches it in multiple ways. First, our survey is more recent in such a quickly evolving field as DASiS, so we address an important set of recent works appeared after their survey. Second, while we organize the DASiS methods according to how domain alignment is achieved similarly to (Toldo *et al.*, 2020a) – namely on *image, feature or output level* – we complement it with different ways of grouping DASiS approaches, namely based on their most important *characteristics*, such as the backbone used for the segmentation network, the type and levels of domain alignments, any complementary techniques used and finally the particularity of each method compared to the others. We report our schema in Table 2.1, which represents one of the core contributions of this monograph. Third, we survey a large set of complementary techniques in Section 2.3 that can help boost the adaptation performance, such as self-training, co-training, self-ensembling and model distillation.

Finally, in Section 2.4 we propose a detailed categorization of some of the *related DA tasks* – such as multi-source, multi-target domain adaptation, domain generalization, source-free adaptation, domain incremental learning, etc. – and survey solutions proposed in the literature to address them. None of the previous surveys has such a comprehensive survey on these related DA tasks, especially what concerns semantic image segmentation.

To complement the above two sections, which represent the core contributions of our monograph, we further provide in Section 3 a list of the datasets and benchmarks typically used to evaluate SiS and DASiS methods, covering the main metrics and discuss different SiS and DASiS evaluation protocols. Furthermore, in Section 4 we propose a short overview of the literature for three tasks strongly related to SiS, namely instance segmentation in Section 4.1, panoptic segmentation in Section 4.2 and medical image segmentation in Section 4.3.

We hope that our monograph, with its comprehensive survey of the main trends in the field of semantic image segmentation, will provide researchers both across academia and in the industry with a solid basis and background to help them develop new methods and foster new research directions.

References

- Ackaouy, A., N. Courty, E. Vallée, O. Commowick, C. Barillot, and F. Galassi. (2020). “Unsupervised Domain Adaptation With Optimal Transport in Multi-Site Segmentation of Multiple Sclerosis Lesions From MRI Data”. *Frontiers in Computational Neuroscience*. 14.
- Adams, R. and L. Bischof. (1994). “Seeded Region Growing”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 16(6): 641–647.
- Ahn, J. and S. Kwak. (2018). “Learning Pixel-level Semantic Affinity with Image-level Supervision for Weakly Supervised Semantic Segmentation”. In: *CVPR*.
- Ahuja, A. S. (2019). “The Impact of Artificial Intelligence in Medicine on the Future Role of the Physician”. *PeerJ*. 7.
- Ali, A., H. Touvron, M. Caron, P. Bojanowski, M. Douze, A. Joulin, I. Laptev, N. Neverova, G. Synnaeve, J. Verbeek, and H. Jegou. (2021). “XCiT: Cross-covariance Image Transformers”. In: *NeurIPS*.
- Alonso, I., A. Sabater, D. Ferstl, L. Montesano, and A. C. Murillo. (2021). “Semi-Supervised Semantic Segmentation with Pixel-Level Contrastive Learning from a Class-wise Memory Bank”. In: *ICCV*.
- Anthimopoulos, M., S. Christodoulidis, L. Ebner, T. Geiser, A. Christe, and S. Mougiakakou. (2019). “Semantic Segmentation of Pathological Lung Tissue with Dilated Fully Convolutional Networks”. *IEEE Journal of Biomedical Health Information*. 23: 714–722.

- Araslanov, N. and S. Roth. (2021). “Self-supervised Augmentation Consistency for Adapting Semantic Segmentation”. In: *CVPR*.
- Arbelaez, P., M. Maire, C. Fowlkes, and J. Malik. (2011). “Contour Detection and Hierarchical Image Segmentation”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 33(5): 898–916.
- Arjovsky, M., L. Bottou, I. Gulrajani, and D. Lopez-Paz. (2020). “Invariant Risk Minimization”. arXiv:1907.02893.
- Arnab, A. and P. H. S. Torr. (2017). “Pixelwise Instance Segmentation with a Dynamically Instantiated Network”. In: *CVPR*.
- Arun, A., C. V. Jawahar, and M. P. Kumar. (2020). “Weakly Supervised Instance Segmentation by Learning Annotation Consistent Instances”. In: *ECCV*.
- Bachmann, R., D. Mizrahi, A. Atanov, and A. Zamir. (2022). “Multi-MAE: Multi-modal Multi-task Masked Autoencoders”. arXiv:2204.01678.
- Badrinarayanan, V., A. Kendall, and R. Cipolla. (2017). “Segnet: a Deep Convolutional Encoder-Decoder Architecture for Image Segmentation”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 39(12): 2481–2495.
- Baevski, A., W.-N. Hsu, Q. Xu, A. Babu, J. Gu, and M. Auli. (2022). “data2vec: A General Framework for Self-supervised Learning in Speech, Vision and Language”. arXiv:2202.03555.
- Balaji, Y., S. Sankaranarayanan, and R. Chellappa. (2019). “MetaReg: Towards Domain Generalization using Meta-Regularization”. In: *NeurIPS*.
- Bao, H., L. Dong, S. Piao, and F. Wei. (2022). “BEiT: BERT Pre-Training of Image Transformers”. In: *ICLR*.
- Bearman, A., O. Russakovsky, V. Ferrari, and L. Fei-Fei. (2016). “What’s the Point: Semantic Segmentation with Point Supervision”. In: *ECCV*.
- Ben-David, S., J. Blitzer, K. Crammer, A. Kulesza, F. Pereira, and J. W. Vaughan. (2010). “A Theory of Learning from Different Domains”. *Machine Learning*. 79(5): 151–175.
- Bengio, Y., J. Louradour, R. Collobert, and J. Weston. (2009). “Curriculum Learning”. In: *ICML*.

- Berman, M., A. R. Triki, and M. B. Blaschko. (2018). “The Lovász-Softmax Loss: A Tractable Surrogate for the Optimization of the Intersection-Over-Union Measure in Neural Networks”. In: *CVPR*.
- Bermúdez-Chacón, R., P. M’arquez-Neila, M. Salzmann, and P. Fua. (2018). “A Domain-adaptive Two-stream U-Net for Electron Microscopy Image Segmentation”. In: *International Symposium on Biomedical Imaging (ISBI)*.
- Blei, D. M., A. Y. Ng, and M. I. Jordan. (2003). “Latent Dirichlet Allocation”. *Journal of Machine Learning Research*. 3: 993–1022.
- Blum, H., P.-E. Sarlin, J. Nieto, R. Siegwart, and C. Cadena. (2019). “Fishscapes: A Benchmark for Safe Semantic Segmentation in Autonomous Driving”. In: *ICCV Workshops*.
- Bolya, D., C. Zhou, F. Xiao, and Y. J. Lee. (2019). “YOLACT: Real-time Instance Segmentation”. In: *ICCV*.
- Borenstein, E. and S. Ullman. (2004). “Learning to Segment”. In: *ECCV*.
- Borgwardt, K., A. Gretton, M. J. Rasch, H.-P. Kriegel, B. Schölkopf, and A. J. Smola. (2006). “Integrating Structured Biological Data by Kernel Maximum Mean Discrepancy”. *Bioinformatics*. 22: 49–57.
- Borne, L., J.-F. Mangin, and D. Rivière. (2019). “Combining 3D U-Net and Bottom-up Geometric Constraints for Automatic Cortical Sulci Recognition”. In: *Medical Imaging with Deep Learning*.
- Borse, S., Y. Wang, Y. Zhang, and F. Porikli. (2021). “InverseForm: A Loss Function for Structured Boundary-Aware Segmentation”. In: *CVPR*.
- Bousmalis, K., N. Silberman, D. Dohan, D. Erhan, and D. Krishnan. (2017). “Unsupervised Pixel-Level Domain Adaptation With Generative Adversarial Networks”. In: *CVPR*.
- Boykov, Y. and M.-P. Jolly. (2001). “Interactive Graph Cuts for Optimal Boundary and Region Segmentation of Objects in N-D Images”. In: *ICCV*.
- Bromley, J., J. W. Bentz, L. Bottou, I. Guyon, Y. LeCun, C. Moore, E. Säckinger, and R. Shah. (1993). “Signature Verification Using a “Siamese” Time Delay Neural Network”. *International Journal of Pattern Recognition and Artificial Intelligence*. 7(04): 669–688.

- Brostow, G. J., J. Fauqueur, and R. Cipolla. (2009). “Semantic Object Classes in Video: a High-definition Ground Truth Database”. *Pattern Recognition Letters*. 30(2): 88–89.
- Bucci, S., A. D’Innocente, Y. Liao, F. M. Carlucci, B. Caputo, and T. Tommasi. (2021). “Self-Supervised Learning Across Domains”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*.
- Buló, S. R., G. Neuhold, and P. Kotschieder. (2017). “Loss Max-Pooling for Semantic Image Segmentation”. In: *CVPR*.
- Byeon, W., T. M. Breuel, F. Raue, and M. Liwicki. (2015). “Scene Labeling with LSTM Recurrent Neural Networks”. In: *CVPR*.
- Cabon, Y., N. Murray, and M. Humenberger. (2020). “Virtual KITTI 2”. arXiv:2001.10773.
- Caesar, H., J. Uijlings, and V. Ferrari. (2018). “COCO-Stuff: Thing and Stuff Classes in Context”. In: *CVPR*.
- Caesar, H., J. R. R. Uijlings, and V. Ferrari. (2015). “Joint Calibration for Semantic Segmentation”. In: *BMVC*.
- Caltagirone, L., M. Bellone, L. Svensson, and M. Wahde. (2019). “LI-DAR-Camera Fusion for Road Detection Using Fully Convolutional Neural Networks”. *Robotics and Autonomous Systems (RAS)*. 111: 125–131.
- Cao, L. and L. Fei-Fei. (2007). “Spatially Coherent Latent Topic Model for Concurrent Object Segmentation and Classification”. In: *ICCV*.
- Cao, Z., L. Ma, M. Long, and J. Wang. (2018). “Partial Adversarial Domain Adaptation”. In: *ECCV*.
- Cao, Z., K. You, M. Long, J. Wang, and Q. Yang. (2019). “Learning to Transfer Examples for Partial Domain Adaptation”. In: *CVPR*.
- Carlucci, F. M., A. D’Innocente, S. Bucci, B. Caputo, and T. Tommasi. (2019). “Domain Generalization by Solving Jigsaw Puzzles”. In: *CVPR*.
- Cermelli, F., D. Fontanel, A. Tavera, M. Ciccone, and B. Caputo. (2022). “Incremental Learning in Semantic Segmentation from Image Labels”. In: *CVPR*.
- Cermelli, F., M. Mancini, S. Rota Bulò, E. Ricci, and B. Caputo. (2020). “Modeling the Background for Incremental Learning in Semantic Segmentation”. In: *CVPR*.

- Cermelli, F., M. Mancini, Y. Xian, Z. Akata, and B. Caputo. (2021). “Prototype-based Incremental Few-Shot Semantic Segmentation”. In: *BMVC*.
- Cesa-Bianchi, N. and G. Lugosi. (2006). *Prediction, Learning, and Games*. Cambridge University Press.
- Cha, S., B. Kim, Y. Yoo, and T. Moon. (2021). “SSUL: Semantic Segmentation with Unknown Label for Exemplar-based Class-Incremental Learning”. In: *NeurIPS*.
- Chan, R., K. Lis, S. Uhlemeyer, H. Blum, S. Honari, R. Siegwart, P. Fua, M. Salzmann, and M. Rottmann. (2021). “SegmentMeIfYouCan: A Benchmark for Anomaly Segmentation”. In: *NeurIPS*.
- Chandra, S. and I. Kokkinos. (2016). “Fast, Exact and Multi-Scale Inference for Semantic Image Segmentation with Deep Gaussian CRFs”. In: *ECCV*.
- Chang, W.-L., H.-P. Wang, W.-H. Peng, and W.-C. Chiu. (2019a). “All about Structure: Adapting Structural Information across Domains for Boosting Semantic Segmentation”. In: *CVPR*.
- Chang, W.-G., T. You, S. Seo, S. Kwak, and B. Han. (2019b). “Domain-Specific Batch Normalization for Unsupervised Domain Adaptation”. In: *CVPR*.
- Chattopadhyay, P., Y. Balaji, and J. Hoffman. (2020). “Learning to Balance Specificity and Invariance for in and out of Domain Generalization”. In: *ECCV*.
- Chaurasia, A. and E. Culurciello. (2017). “LinkNet: Exploiting Encoder Representations for Efficient Semantic Segmentation”. In: *VCIP*.
- Chen, H., K. Sun, Z. Tian, C. Shen, Y. Huang, and Y. Yan. (2020a). “BlendMask: Top-Down Meets Bottom-Up for Instance Segmentation”. In: *CVPR*.
- Chen, Y.-H., W.-Y. Chen, Y.-T. Chen, B.-C. Tsai, Y.-C. F. Wang, and M. Sun. (2017a). “No More Discrimination: Cross City Adaptation of Road Scene Segmenters”. In: *ICCV*.
- Chen, L.-C., A. Hermans, G. Papandreou, F. Schroff, P. Wang, and H. Adam. (2018a). “MaskLab: Instance Segmentation by Refining Object Detection with Semantic and Direction Features”. In: *CVPR*.

- Chen, L.-C., G. Papandreou, I. Kokkinos, K. Murphy, and A. L. Yuille. (2017b). “Deeplab: Semantic Image Segmentation with Deep Convolutional Nets, Atrous Convolution, and Fully Connected CRFs”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 40(4): 834–848.
- Chen, L.-C., G. Papandreou, F. Schroff, and H. Adam. (2017c). “Rethinking Atrous Convolution for Semantic Image Segmentation”. arXiv:1706.05587.
- Chen, L.-C., Y. Yang, J. Wang, W. Xu, and A. L. Yuille. (2016). “Attention to Scale: Scale-aware Semantic Image Segmentation”. In: *CVPR*.
- Chen, L.-C., Y. Zhu, G. Papandreou, F. Schroff, and H. Adam. (2018b). “Encoder-Decoder with Atrous Separable Convolution for Semantic Image Segmentation”. In: *ECCV*.
- Chen, M., H. Xue, and D. Cai. (2019a). “Domain Adaptation for Semantic Segmentation with Maximum Squares Loss”. In: *ICCV*.
- Chen, R., Y. Rong, S. Guo, J. Han, F. Sun, T. Xu, and W. Huang. (2022). “Smoothing Matters: Momentum Transformer for Domain Adaptive Semantic Segmentation”. arXiv:2203.07988.
- Chen, S., X. Jia, J. He, Y. Shi, and J. Liu. (2021a). “Semi-Supervised Domain Adaptation Based on Dual-Level Domain Mixing for Semantic Segmentation”. In: *CVPR*.
- Chen, T., S. Kornblith, M. Norouzi, and G. Hinton. (2020b). “A Simple Framework for Contrastive Learning of Visual Representations”. In: *ICML*.
- Chen, Y.-T., X. Liu, and M.-H. Yang. (2015). “Multi-instance Object Segmentation with Occlusion Handling”. In: *CVPR*.
- Chen, X., A. Jain, A. Gupta, and L. S. Davis. (2011). “Piecing Together the Segmentation Jigsaw using Context”. In: *CVPR*.
- Chen, X., Y. Yuan, G. Zeng, and J. Wang. (2021b). “Semi-Supervised Semantic Segmentation with Cross Pseudo Supervision”. In: *CVPR*.
- Chen, Y., W. Li, X. Chen, and L. Van Gool. (2019b). “Learning semantic segmentation from synthetic data: A geometrically guided input-output adaptation approach”. In: *CVPR*.
- Chen, Y., W. Li, and L. Van Gool. (2018c). “Road: Reality Oriented Adaptation for Semantic Segmentation of Urban Scenes”. In: *CVPR*.

- Chen, Y.-C., Y.-Y. Lin, M.-H. Yang, and J.-B. Huang. (2019c). “CrDoCo: Pixel-level Domain Transfer with Cross-Domain Consistency”. In: *CVPR*.
- Chen, Z., V. Badrinarayanan, C.-Y. Lee, and A. Rabinovich. (2018d). “GradNorm: Gradient Normalization for Adaptive Loss Balancing in Deep Multitask Networks”. In: *ICML*.
- Chen, Z., J. Zhuang, X. Liang, and L. Lin. (2019d). “Blending-target Domain Adaptation by Adversarial Meta-Adaptation Networks”. In: *CVPR*.
- Cheng, B., M. Collins, Y. Zhu, T. Liu, T. S. Huang, H. Adam, and L.-C. Chen. (2020). “Panoptic-DeepLab: A Simple, Strong, and Fast Baseline for Bottom-Up Panoptic Segmentation”. In: *CVPR*.
- Cheng, B., I. Misra, A. G. Schwing, A. Kirillov, and R. Girdhar. (2022). “Masked-attention Mask Transformer for Universal Image Segmentation”. In: *CVPR*.
- Cheng, Y., F. Wei, J. Bao, D. Chen, F. Wen, and W. Zhang. (2021). “Dual Path Learning for Domain Adaptation of Semantic Segmentation”. In: *ICCV*.
- Chidlovskii, B., S. Clinchant, and G. Csurka. (2016). “Domain Adaptation in the Absence of Source Domain Data”. In: *PKDD*.
- Choi, J., T. Kim, and C. Kim. (2019). “Self-Ensembling with GAN-based Data Augmentation for Domain Adaptation in Semantic Segmentation”. In: *ICCV*.
- Choi, S., S. Jung, H. Yun, J. Kim, S. Kim, and J. Choo. (2021). “RobustNet: Improving Domain Generalization in Urban-Scene Segmentation via Instance Selective Whitening”. In: *CVPR*.
- Chu, X., Z. Tian, Y. Wang, B. Zhang, H. Ren, X. Wei, H. Xia, and C. Shen. (2021). “Twins: Revisiting the Design of Spatial Attention in Vision Transformers”. In: *NeurIPS*.
- Clinchant, S., J.-M. Renders, and G. Csurka. (2007). “XRCE’s Participation to ImageCLEF”. In: *CLEF Online Working Notes*.
- Comanicu, D. and P. Meer. (2002). “Mean Shift: A Robust Approach Toward Feature Space Analysis”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 24(5): 603–619.

- Corbière, C., N. Thome, A. Saporta, T.-H. Vu, M. Cord, and P. Perez. (2021). “Confidence Estimation via Auxiliary Models”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*.
- Cordts, M., M. Omran, S. Ramos, T. Rehfeld, M. Enzweiler, R. Benenson, U. Franke, S. Roth, and B. Schiele. (2016). “The Cityscapes Dataset for Semantic Urban Scene Understanding”. In: *CVPR*.
- Cortes, C., M. Mohri, A. T. Suresh, and N. Zhang. (2021). “A Discriminative Technique for Multiple-Source Adaptation”. In: *ICML*.
- Crammer, K., M. Kearns, and J. Wortman. (2008). “Learning from Multiple Sources”. *Journal of Machine Learning Research*. 9.
- Csurka, G. (2017). “A Comprehensive Survey on Domain Adaptation for Visual Applications”. In: *Domain Adaptation in Computer Vision Applications*. Ed. by G. Csurka. *Advances in Computer Vision and Pattern Recognition*. Springer. 1–35.
- Csurka, G. (2020). “Deep Visual Domain Adaptation”. arXiv:2012.14176.
- Csurka, G., F. Baradel, B. Chidlovskii, and S. Clinchant. (2017). “Discrepancy-Based Networks for Unsupervised Domain Adaptation: A Comparative Study”. In: *ICCV Workshop on Transferring and Adapting Source Knowledge in Computer Vision (TASK-CV)*.
- Csurka, G., C. R. Dance, L. Fan, J. Willamowski, and C. Bray. (2004). “Visual Categorization with Bags of Keypoints”. In: *ECCV Workshop on Statistical Learning in Computer Vision (SLCV)*.
- Csurka, G., D. Larlus, and F. Perronin. (2013). “What is a Good Evaluation Measure for Semantic Segmentation?” In: *BMVC*.
- Csurka, G. and F. Perronin. (2011). “An Efficient Approach to Semantic Segmentation”. *International Journal of Computer Vision (IJCV)*. 95: 198–212.
- Dai, J., K. He, Y. Li, S. Ren, and J. Sun. (2016). “Instance-sensitive Fully Convolutional Networks”. In: *ECCV*.
- Dai, J., K. He, and J. Sun. (2015). “BoxSup: Exploiting Bounding Boxes to Supervise Convolutional Networks for Semantic Segmentation”. In: *ICCV*.
- de Geus, D., P. Meletis, and G. Dubbelman. (2018). “Panoptic Segmentation with a Joint Semantic and Instance Segmentation Network”. arXiv:1809.02110.

- DeChicchis, J. (2020). “Semantic Understanding for Augmented Reality and Its Applications”. *Tech. rep.* Duke University.
- Devlin, J., M.-W. Chang, K. Lee, and K. Toutanova. (2019). “BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding”. In: *NAACL HLT*.
- Di, S., H. Zhang, C.-G. Li, X. Mei, D. Prokhorov, and H. Ling. (2018). “Cross-Domain Traffic Scene Understanding: A Dense Correspondence-Based Transfer Learning Approach”. *IEEE Transactions on Intelligent Transportation Systems*. 19(3): 745–757.
- Ding, H., C. Liu, S. Wang, and X. Jiang. (2021). “Vision-Language Transformer and Query Generation for Referring Segmentation”. In: *CVPR*.
- Dosovitskiy, A., L. Beyer, A. Kolesnikov, D. Weissenborn, X. Zhai, T. Unterthiner, M. Dehghani, M. Minderer, G. Heigold, S. Gelly, J. Uszkoreit, and N. Houlsby. (2021). “An Image is Worth 16x16 Words: Transformers for Image Recognition at Scale”. In: *ICLR*.
- Dou, Q., C. Ouyang, C. Chen, H. Chen, and P.-A. Heng. (2018). “Un-supervised Cross-Modality Domain Adaptation of ConvNets for Biomedical Image Segmentations with Adversarial Loss”. In: *IJCAI*.
- Douillard, A., Y. Chen, A. Dapogny, and M. Cord. (2021). “PLOP: Learning Without Forgetting for Continual Semantic Segmentation”. In: *CVPR*.
- Du, L., J. Tan, H. Yang, J. Feng, X. Xue, Q. Zheng, X. Ye, and X. Zhang. (2019). “SSF-DAN: Separated Semantic Feature Based Domain Adaptation Network for Semantic Segmentation”. In: *ICCV*.
- Durand, T., T. Mordan, N. Thome, and M. Cord. (2017). “WILD-CAT: Weakly Supervised Learning of Deep ConvNets for Image Classification, Pointwise Localization and Segmentation”. In: *CVPR*.
- Edupuganti, V. G., A. Chawla, and A. Kale. (2017). “Automatic Optic Disk and Cup Segmentation of Fundus Images Using Deep Learning”. In: *ICIP*.
- Eigen, D. and R. Fergus. (2015). “Predicting Depth, Surface Normals and Semantic Labels with a Common Multi-Scale Convolutional Architecture”. In: *CVPR*.

- Everingham, M., L. Van Gool, C. Williams, J. Winn, and A. Zisserman. (2010). “The Pascal Visual Object Classes (VOC) Challenge”. *International Journal of Computer Vision (IJCV)*. 88: 303–338.
- Fan, R., Q. Hou, M.-M. Cheng, G. Yu, R. R. Martin, and S.-M. Hu. (2018). “Associating Inter-Image Salient Instances for Weakly Supervised Semantic Segmentation”. In: *ECCV*.
- Fang, K., Y. Bai, S. Hinterstoisser, S. Savarese, and M. Kalakrishnan. (2018). “Multi-Task Domain Adaptation for Deep Learning of Instance Grasping from Simulation”. In: *ICRA*.
- Fang, Y., L. Dong, H. Bao, X. Wang, and F. Wei. (2022). “Corrupted Image Modeling for Self-Supervised Visual Pre-Training”. arXiv:2202.03382.
- Farabet, C., C. Couprie, L. Najman, and Y. LeCun. (2012). “Scene Parsing with Multiscale Feature Learning, Purity Trees, and Optimal Covers”. In: *ICML*.
- Farabet, C., C. Couprie, L. Najman, and Y. LeCun. (2013). “Learning Hierarchical Features for Scene Labeling”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 35(8): 1915–1929.
- Feng, D., C. Haase-Schütz, L. Rosenbaum, H. Hertlein, C. Glaeser, F. Timm, W. Wiesbeck, and K. Dietmayer. (2021). “Deep Multi-Modal Object Detection and Semantic Segmentation for Autonomous Driving: Datasets, Methods, and Challenges”. *IEEE Intelligent Transportation Systems Conference*. 22(3): 1341–1360.
- Feng, Z., Q. Zhou, X. Tan, G. Cheng, X. Lu, J. Shi, and L. Ma. (2020). “DMT: Dynamic Mutual Training for Semi-Supervised Learning”. arXiv:2004.08514.
- Fourure, D., R. Emonet, E. Fromont, D. Muselet, A. Tremeau, and C. Wolf. (2017). “Residual Conv-Deconv Grid Network for Semantic Segmentation”. In: *BMVC*.
- French, G., T. Aila, S. Laine, M. Mackiewicz, and G. Finlayson. (2019). “Semi-supervised Semantic Segmentation Needs Strong, High-dimensional Perturbations”. In: *BMVC*.
- French, G., S. Laine, T. Aila, and M. Mackiewicz. (2020). “Semi-supervised Semantic Segmentation Needs Strong, Varied Perturbations”. In: *BMVC*.

- French, G., M. Mackiewicz, and M. Fisher. (2018). “Self-ensembling for Visual Domain Adaptation”. In: *ICLR*.
- Fu, B., Z. Cao, M. Long, and J. Wang. (2020). “Learning to Detect Open Classes for Universal Domain Adaptation”. In: *ECCV*.
- Fu, J., J. Liu, H. Tian, Y. Li, Y. Bao, Z. Fang, and H. Lu. (2019a). “Dual Attention Network for Scene Segmentation”. In: *CVPR*.
- Fu, J., J. Liu, Y. Wang, and H. Lu. (2017). “Densely Connected Deconvolutional Network for Semantic Segmentation”. In: *ICIP*.
- Fu, J., J. Liu, Y. Wang, J. Zhou, C. Wang, and H. Lu. (2019b). “Stacked Deconvolutional Network for Semantic Segmentation”. *IEEE Transactions on Image Processing (TIP)*.
- Galleguillos, C., A. Rabinovich, A. Rabinovich, and S. Belongie. (2008). “Weakly Supervised Object Localization with Stable Segmentations”. In: *ECCV*.
- Ganin, Y., E. Ustinova, P. Ajakan Hana And Germain, H. Larochelle, F. Laviolette, M. Marchand, and V. S. Lempitsky. (2016). “Domain-Adversarial Training of Neural Networks”. *Journal of Machine Learning Research*.
- Gao, N., Y. Shan, Y. Wang, X. Zhao, Y. Yu, M. Yang, and K. Huang. (2019). “SSAP: Single-Shot Instance Segmentation With Affinity Pyramid”. In: *ICCV*.
- Gao, T., W. Wei, Z. Cai, Z. Fan, S. Xie, X. Wang, and Q. Yu. (2022). “CI-Net: Contextual Information for Joint Semantic Segmentation and Depth Estimation”. *Applied Intelligence*. Open access.
- Gao, W., F. Wan, X. Pan, Z. Peng, Q. Tian, Z. Han, B. Zhou, and Q. Ye. (2021). “TS-CAM: Token Semantic Coupled Attention Map for Weakly Supervised Object Localization”. In: *ICCV*.
- Garg, P., R. Saluja, V. N. Balasubramanian, C. Arora, A. Subramanian, and C. Jawahar. (2022). “Multi-Domain Incremental Learning for Semantic Segmentation”. In: *WACV*.
- Garg, S., N. Sünderhauf, F. Dayoub, D. Morrison, A. Cosgun, G. Carneiro, Q. Wu, T.-J. Chin, I. Reid, S. Gould, P. Corke, and M. Milford. (2020). “Semantics for Robotic Mapping, Perception and Interaction: A Survey”. *Foundations and Trends® in Robotics*. 8(1-2): 1–224.

- Gatta, C., A. Romero, and J. van de Veijer. (2014). “Unrolling Loopy Top-Down Semantic Feedback in Convolutional Deep Networks”. In: *CVPR Workshops*.
- Geiger, A., P. Lenz, and R. Urtasun. (2012). “Are We Ready for Autonomous Driving? the KITTI Vision Benchmark Suite”. In: *CVPR*.
- Ghiasi, G. and C. C. Fowlkes. (2016). “Laplacian Pyramid Reconstruction and Refinement for Semantic Segmentation”. In: *ECCV*.
- Gholami, B., P. Sahu, O. Rudovic, K. Bousmalis, and V. Pavlovic. (2020). “Unsupervised Multi-Target Domain Adaptation: An Information Theoretic Approach”. *IEEE Transactions on Image Processing (TIP)*. 29(1): 3993–4002.
- Gidaris, S., P. Singh, and N. Komodakis. (2018). “Unsupervised Representation Learning by Predicting Image Rotations”. In: *ICLR*.
- Gilmer, J., N. Ford, N. Carlini, and E. Cubuk. (2019). “Adversarial Examples Are a Natural Consequence of Test Error in Noise”. In: *ICML*.
- Girshick, R., J. Donahue, T. Darrell, and J. Malik. (2014). “Rich Feature Hierarchies for Accurate Object Detection and Semantic Segmentation”. In: *CVPR*.
- Gonfous, J. M., X. Boix, J. van de Weijer, A. D. Bagdanov, J. Serrat, and J. González. (2010). “Harmony Potentials for Joint Classification and Segmentation”. In: *CVPR*.
- Gong, R., Y. Chen, D. P. Paudel, Y. Li, A. Chhatkuli, W. Li, D. Dai, and L. Van Gool. (2021a). “Cluster, Split, Fuse, and Update: Meta-Learning for Open Compound Domain Adaptive Semantic Segmentation”. In: *CVPR*.
- Gong, R., D. Dai, Y. Chen, W. Li, and L. Van Gool. (2021b). “mDALU: Multi-Source Domain Adaptation and Label Unification with Partial Datasets”. In: *ICCV*.
- Gong, R., W. Li, Y. Chen, and L. Van Gool. (2019). “DLOW: Domain Flow for Adaptation and Generalization”. In: *CVPR*.
- Goodfellow, I., J. Pouget-Abadie, M. Mirza, B. Xu, D. Warde-Farley, S. Ozair, A. Courville, and Y. Bengio. (2014). “Generative Adversarial Nets”. In: *NeurIPS*.

- Gopalan, R., R. Li, V. M. Patel, and R. Chellappa. (2015). “Domain Adaptation for Visual Recognition”. *Foundations and Trends® in Computer Graphics and Vision*. 8(4): 285–378.
- Gould, S., R. Fulton, and D. Koller. (2009). “Decomposing a Scene into Geometric and Semantically Consistent Regions”. In: *ICCV*.
- Gould, S., J. Rodgers, D. Cohen, G. Elidan, and D. Koller. (2008). “Multi-Class Segmentation with Relative Location Prior”. *International Journal of Computer Vision (IJCV)*. 80: 300–316.
- Grandvalet, Y. and Y. Bengio. (2004). “Semi-supervised Learning by Entropy Minimization”. In: *NeurIPS*.
- Grangier, D., L. Bottou, and R. Collobert. (2009). “Deep Convolutional Networks for Scene Parsing”. In: *ICML*.
- Greenspan, H., B. van Ginneken, and R. M. Summers. (2016). “Guest Editorial Deep Learning in Medical Imaging: Overview and Future Promise of an Exciting New Technique”. *IEEE Transactions on Medical Imaging*. 35(5): 1153–1159.
- Gretton, A., A. Smola, J. Huang, M. Schmittfull, K. Borgwardt, and B. Schölkopf. (2009). “Covariate Shift by Kernel Mean Matching”. In: *Dataset Shift in Machine Learning*. Ed. by J. Quiñero-Candela, M. Sugiyama, A. Schwaighofer, and N. D. Lawrence. The MIT Press.
- Gu, C., J. J. Lim, P. Arbeláez, and J. Malik. (2009). “Recognition using Regions”. In: *CVPR*.
- Gu, Q., Q. Zhou, M. Xu, Z. Feng, G. Cheng, X. Lu, J. Shi, and L. Ma. (2021). “PIT: Position-Invariant Transform for Cross-FoV Domain Adaptation”. In: *ICCV*.
- Guizilini, V., J. Li, R. Ambrus, and A. Gaidon. (2021). “Geometric Unsupervised Domain Adaptation for Semantic Segmentation”. In: *ICCV*.
- Guo, R., D. Niu, L. Qu, and Z. Li. (2021a). “SOTR: Segmenting Objects With Transformers”. In: *ICCV*.
- Guo, X., C. Yang, B. Li, and Y. Yuan. (2021b). “MetaCorrection: Domain-Aware Meta Loss Correction for Unsupervised Domain Adaptation in Semantic Segmentation”. In: *CVPR*.
- Ha, Q., K. Watanabe, T. Karasawa, Y. Ushiku, and T. Harada. (2017). “MFNet: Towards Real-time Semantic Segmentation for Autonomous Vehicles with Multi-spectral Scenes”. In: *IROS*.

- Han, Z., B. Wei, A. Mercado, S. Leung, and S. Li. (2018). “Spine-GAN: Semantic Segmentation of Multiple Spinal Structures”. *Medical Image Analyses*. 79: 2379–2391.
- Hariharan, B., P. Arbeláez, R. Girshick, and J. Malik. (2014). “Simultaneous Detection and Segmentation”. In: *ECCV*.
- Hariharan, B., P. Arbeláez, R. Girshick, and J. Malik. (2015). “Hypercolumns for Object Segmentation and Fine-grained Localization”. In: *CVPR*.
- Hassani, A., S. Walton, J. Li, S. Li, and H. Shi. (2022). “Neighborhood Attention Transformer”. arXiv:2204.07143.
- Hazirbas, C., L. Ma, C. Domokos, and D. Cremers. (2016). “FuseNet: Incorporating Depth into Semantic Segmentation via Fusion-Based CNN Architecture”. In: *ACCV*.
- He, J., X. Jia, S. Chen, and J. Liu. (2021a). “Multi-Source Domain Adaptation With Collaborative Learning for Semantic Segmentation”. In: *CVPR*.
- He, J., Z. Deng, L. Zhou, Y. Wang, and Y. Qiao. (2019). “Adaptive Pyramid Context Network for Semantic Segmentation”. In: *CVPR*.
- He, K., X. Chen, S. Xie, Y. Li, P. Dollár, and R. Girshick. (2022). “Masked Autoencoders are Scalable Vision Learners”. In: *CVPR*.
- He, K., G. Gkioxari, P. Dollár, and R. Girshick. (2017). “Mask R-CNN”. In: *ICCV*.
- He, K., X. Zhang, S. Ren, and J. Sun. (2014). “Spatial Pyramid Pooling in Deep Convolutional Networks for Visual Recognition”. In: *ECCV*.
- He, L., J. Lu, G. Wang, S. Song, and J. Zhou. (2021b). “SOSD-Net: Joint Semantic Object Segmentation and Depth Estimation from Monocular images”. *Neurocomputing*. 440: 251–263.
- He, R., J. Yang, and X. Qi. (2021c). “Re-distributing Biased Pseudo Labels for Semi-supervised Semantic Segmentation: A Baseline Investigation”. In: *CVPR*.
- He, X. and R. S. Zemel. (2009). “Learning Hybrid Models for Image Annotation with Partially Labeled Data”. In: *NeurIPS*.
- He, X., R. S. Zemel, and M. Á. Carreira-Perpiñán. (2004). “Multiscale Conditional Random Fields for Image Labeling”. In: *CVPR*.
- He, X., R. S. Zemel, and D. Ray. (2006). “Learning and Incorporating Top-Down Cues in Image Segmentation”. In: *ECCV*.

- Hendrycks, D. and T. Dietterich. (2019). “Benchmarking Neural Network Robustness to Common Corruptions and Perturbations”. In: *ICLR*.
- Hendrycks, D. and K. Gimpel. (2017). “A Baseline for Detecting Misclassified and Out-of-Distribution Examples in Neural Networks”. In: *ICLR*.
- Hinton, G., O. Vinyals, and J. Dean. (2015). “Distilling the Knowledge in a Neural Network”. arXiv:1503.02531.
- Hoffman, J., M. Mohri, and N. Zhang. (2018a). “Algorithms and Theory for Multiple-Source Adaptation”. In: *NeurIPS*.
- Hoffman, J., E. Tzeng, T. Park, J.-Y. Zhu, P. Isola, K. Saenko, A. A. Efros, and T. Darrel. (2018b). “CyCADA: Cycle-Consistent Adversarial Domain Adaptation”. In: *ICML*.
- Hoffman, J., D. Wang, F. Yu, and T. Darrell. (2016). “FCNs in the Wild: Pixel-level Adversarial and Constraint-based Adaptation”. arXiv:1612.02649.
- Hofmann, T. (2001). “Unsupervised Learning by Probabilistic Latent Semantic Analysis”. *Machine Learning*. 42(1-2): 177–196.
- Hofmarcher, M., T. Unterthiner, J. Arjona-Medina, G. Klambauer, S. Hochreiter, and B. Nessler. (2019). “Visual Scene Understanding for Autonomous Driving Using Semantic Segmentation”. In: *Explainable AI: Interpreting, Explaining and Visualizing Deep Learning*. Ed. by W. Samek, G. Montavon, A. Vedaldi, L. Hansen, and K.-R. Müller. *Advances in Computer Vision and Pattern Recognition*. Springer. 285–296.
- Hoi, S. C., D. Sahoo, J. Lu, and P. Zhao. (2018). “Online Learning: A Comprehensive Survey”. arXiv:1802.02871.
- Hong, S., H. Noh, and B. Han. (2015). “Decoupled Deep Neural Network for Semi-supervised Semantic Segmentation”. In: *NeurIPS*.
- Hong, S., J. Oh, B. Han, and H. Lee. (2016). “Learning Transferrable Knowledge for Semantic Segmentation with Deep Convolutional Neural Network”. In: *CVPR*.
- Hong, S., D. Yeo, S. Kwak, H. Lee, and B. Han. (2017). “Weakly Supervised Semantic Segmentation using Web-Crawled Video”. In: *CVPR*.

- Hong, W., Z. Wang, M. Yang, and J. Yuan. (2018a). “Conditional Generative Adversarial Network for Structured Domain Adaptation”. In: *CVPR*.
- Hong, Z.-W., Y.-M. Chen, H.-K. Yang, S.-Y. Su, T.-Y. Shann, Y. H. Chang, B. Hsi-Lin Ho, C.-C. Tu, T.-C. Hsiao, H.-W. Hsiao, S.-P. Lai, Y.-C. Chang, and C.-Y. Lee. (2018b). “Virtual-to-Real: Learning to Control in Visual Semantic Segmentation”. In: *IJCAI*.
- Hoyer, L., D. Dai, and L. Van Gool. (2022). “DAFormer: Improving Network Architectures and Training Strategies for Domain-Adaptive Semantic Segmentation”. In: *CVPR*.
- Hu, H., F. Wei, H. Hu, Q. Ye, J. Cui, and L. Wang. (2021). “Semi-Supervised Semantic Segmentation via Adaptive Equalization Learning”. In: *NeurIPS*.
- Hu, R., M. Rohrbach, and T. Darrell. (2016). “Segmentation from Natural Language Expressions”. In: *ECCV*.
- Hu, R., D. Larlus, and G. Csurka. (2012). “On the Use of Regions for Semantic Image Segmentation”. In: *ICCVGIP*.
- Hu, X., C.-W. Fu, L. Zhu, and P.-A. Heng. (2019a). “Depth-Attentional Features for Single-Image Rain Removal”. In: *CVPR*.
- Hu, X., K. Yang, L. Fei, and K. Wang. (2019b). “ACNET: Attention Based Network to Exploit Complementary Features for RGBD Semantic Segmentation”. In: *ICIP*.
- Huang, H., Q. Huang, and P. Krähenbühl. (2018a). “Domain Transfer Through Deep Activation Matching”. In: *ECCV*.
- Huang, J., D. Guan, A. Xiao, and S. Lu. (2021). “RDA: Robust Domain Adaptation via Fourier Adversarial Attacking”. In: *ICCV*.
- Huang, J., S. Lu, D. Guan, and X. Zhang. (2020a). “Contextual-Relation Consistent Domain Adaptation for Semantic Segmentation”. In: *ECCV*.
- Huang, S. and S. Belongie. (2017). “Arbitrary Style Transfer in Real-time with Adaptive Instance Normalization”. In: *ICCV*.
- Huang, Z., H. Mao, N. Jiang, and X. Wang. (2020b). “DAPR-Net: Domain Adaptive Predicting-refinement Network for Retinal Vessel Segmentation”. In: *MICCAI Workshops*.

- Huang, Z., X. Wang, J. Wang, W. Liu, and J. Wang. (2018b). “Weakly-Supervised Semantic Segmentation Network with Deep Seeded Region Growing”. In: *CVPR*.
- Hung, W.-C., Y.-H. Tsai, Y.-T. Liou, Y.-Y. Lin, and M.-H. Yang. (2018). “Adversarial Learning for Semi-supervised Semantic Segmentation”. In: *BMVC*.
- Ioffe, S. (2021). “Batch Renormalization: Towards Reducing Minibatch Dependence in Batch-Normalized Models”. In: *NeurIPS*.
- Ioffe, S. and C. Szegedy. (2015). “Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariate Shift”. In: *ICML*.
- Isobe, T., X. Jia, S. Chen, J. He, Y. Shi, J. Liu, H. Lu, and S. Wang. (2021). “Multi-Target Domain Adaptation with Collaborative Consistency Learning”. In: *CVPR*.
- Jain, J., A. Singh, N. Orlov, Z. Huang, J. Li, S. Walton, and H. Shi. (2021). “SeMask: Semantically Masked Transformers for Semantic Segmentation”. arXiv:2112.12782.
- Jan, E. and X. Chen, eds. (2020). *Computer-Aided Oral and Maxillofacial Surgery: Developments, Applications, and Future Perspectives*. Academic Press.
- Jaritz, M., R. de Charette, E. Wirbel, X. Perrotton, and F. Nashashibi. (2018). “Sparse and Dense Data with CNNs: Depth Completion and Semantic Segmentation”. In: *3DV*.
- Jesson, A., N. Guizard, S. H. Ghalehjegh, D. Goblot, F. Soudan, and N. Chapados. (2017). “CASED: Curriculum Adaptive Sampling for Extreme Data Imbalance”. In: *MICCAI*.
- Ji, Z. and O. Veksler. (2021). “Weakly Supervised Semantic Segmentation: From Box to Tag and Back”. In: *BMVC*.
- Jiang, J., Y.-C. Hu, N. Tyagi, P. Zhang, A. Rimmer, G. S. Mageras, J. O. Deasy, and H. Veeraraghavan. (2018). “Tumor-Aware, Adversarial Domain Adaptation from CT to MRI for Lung Cancer Segmentation”. In: *MICCAI*.
- Jiao, J., Y. Cao, Y. Song, and R. Lau. (2018). “Look Deeper into Depth: Monocular Depth Estimation with Semantic Booster and Attention-Driven Loss”. In: *ECCV*.

- Jin, B., M. V. Ortiz Segovia, and S. Susstrunk. (2017). “Webly Supervised Semantic Segmentation”. In: *CVPR*.
- Jin, X., C. Lan, W. Zeng, and Z. Chen. (2020). “Style Normalization and Restitution for Domain Generalization and Adaptation”. arXiv:2101.00588.
- Jing, T., H. Liu, and Z. Ding. (2021). “Towards Novel Target Discovery Through Open-Set Domain Adaptation”. In: *ICCV*.
- Jurie, F. and B. Triggs. (2005). “Creating Efficient Codebooks for Visual Recognition”. In: *ICCV*.
- Kamann, C. and C. Rother. (2020). “Benchmarking the Robustness of Semantic Segmentation Models”. In: *CVPR*.
- Kato, Z. and J. Zerubia. (2012). “Markov Random Fields in Image Segmentation”. *Foundations and Trends® in Signal Processing*. 5(1-2): 1–155.
- Kendall, A. and Y. Gal. (2017). “What Uncertainties Do We Need in Bayesian Deep Learning for Computer Vision?” In: *NeurIPS*.
- Kendall, A., Y. Gal, and R. Cipolla. (2018). “Multi-Task Learning Using Uncertainty to Weigh Losses for Scene Geometry and Semantics”. In: *CVPR*.
- Khan, S. H., M. Hayat, M. Bennamoun, F. A. Sohel, and R. Togneri. (2018). “Cost-Sensitive Learning of Deep Feature Representations from Imbalanced Data”. *IEEE Transactions on Neural Networks and Learning Systems*. 29(8): 3573–3587.
- Khoreva, A., R. Benenson, J. Hosang, M. Hein, and B. Schiele. (2017). “Simple does it: Weakly Supervised Instance and Semantic Segmentation”. In: *CVPR*.
- Kim, D.-K., D. Maturana, M. Uenoyama, and S. Scherer. (2018). “Season-Invariant Semantic Segmentation with a Deep Multimodal Network”. In: *Field and Service Robotics*. Ed. by M. Hutter and R. Siegwart. *Advanced Robotics*. Springer. 255–270.
- Kim, M. and H. Byun. (2020). “Learning Texture Invariant Representation for Domain Adaptation of Semantic Segmentation”. In: *CVPR*.
- King Jr., B. F. (2018). “Artificial Intelligence and Radiology: What will the Future Hold?” *Journal of the American College of Radiology*. 15(3 Part B): 501–503.

- Kirillov, A., K. He, R. Girshick, C. Rother, and P. Dollár. (2019a). “Panoptic Feature Pyramid Networks”. In: *CVPR*.
- Kirillov, A., K. He, R. Girshick, C. Rother, and P. Dollár. (2019b). “Panoptic Segmentation”. In: *CVPR*.
- Kirillov, A., E. Levinkov, B. Andres, B. Savchynskyy, and C. Rother. (2017). “InstanceCut: from Edges to Instances with MultiCut”. In: *CVPR*.
- Kohli, P., L. Ladický, and P. H. Torr. (2009). “Robust Higher Order Potentials for Enforcing Label Consistency”. *International Journal of Computer Vision (IJCV)*. 82: 302–324.
- Kolesnikov, A. and C. H. Lampert. (2016). “Seed, Expand and Constrain: Three Principles for Weakly-Supervised Image Segmentation”. In: *ECCV*.
- Kothandaraman, D., A. Nambiar, and A. Mittal. (2021). “Domain Adaptive Knowledge Distillation for Driving Scene Semantic Segmentation”. In: *WACV*.
- Kouw, W. M. and M. Loog. (2021). “A Review of Domain Adaptation without Target Labels”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 43(3): 766–785.
- Krähenbühl, P. and V. Koltun. (2011). “Efficient Inference in Fully Connected CRFs with Gaussian Edge Potentials”. In: *NeurIPS*.
- Kulharia, V., S. Chandra, A. Agrawal, P. Torr, and A. Tyagi. (2020). “Box2Seg: Attention Weighted Loss and Discriminative Feature Learning for Weakly Supervised Segmentation”. In: *ECCV*.
- Kumar, M. P., P. H. S. Torr, and A. Zisserman. (2005). “Obj Cut”. In: *CVPR*.
- Kumar, M. P., H. Turki, D. Preston, and D. Koller. (2011). “Learning Specific-Class Segmentation from Diverse Data”. In: *ICCV*.
- Kumar, S. and M. Hebert. (2005). “A Hierarchical Field Framework for Unified Context-Based Classification”. In: *ICCV*.
- Kundu, J. N., A. Kulkarni, A. Singh, V. Jampani, and R. Babu. (2021). “Generalize Then Adapt: Source-Free Domain Adaptive Semantic Segmentation”. In: *ICCV*.
- Kurmi, V. K., V. K. Subramanian, and V. P. Namboodiri. (2021). “Domain Impression: A Source Data Free Domain Adaptation Method”. In: *WACV*.

- Kwak, S., S. Hong, and B. Han. (2017). “Weakly Supervised Semantic Segmentation Using Superpixel Pooling Network”. In: *AAAI*.
- Ladický, L., C. Russell, P. Kohli, and P. H. S. Torr. (2009). “Associative Hierarchical CRFs for Object Class Image Segmentation”. In: *ICCV*.
- Lai, X., Z. Tian, L. Jiang, S. Liu, H. Zhao, L. Wang, and J. Jia. (2021). “Semi-supervised Semantic Segmentation with Directional Context-aware Consistency”. In: *CVPR*.
- Laine, S. and T. Aila. (2016). “Temporal Ensembling for Semisupervised Learning”. arXiv:1610.02242.
- Lan, S., Z. Yu, C. Choy, S. Radhakrishnan, G. Liu, Y. Zhu, L. S. Davis, and A. Anandkumar. (2021). “DiscoBox: Weakly Supervised Instance Segmentation and Semantic Correspondence From Box Supervision”. In: *ICCV*.
- Larlus, D., J. Verbeek, and F. Jurie. (2010). “Category Level Object Segmentation by Combining Bag-of-Words Models with Dirichlet Processes and Random Fields”. *International Journal of Computer Vision (IJCV)*. 88: 238–253.
- Lee, C.-Y., T. Batra, M. H. Baig, and D. Ulbricht. (2019a). “Sliced Wasserstein Discrepancy for Unsupervised Domain Adaptation”. In: *CVPR*.
- Lee, D.-H. (2013). “Pseudo-Label : The Simple and Efficient Semi-Supervised Learning Method for Deep Neural Networks”. In: *ICML Workshops*.
- Lee, J., E. Kim, S. Lee, J. Lee, and S. Yoon. (2019b). “FickleNet: Weakly and Semi-supervised Semantic Image Segmentation using Stochastic Inference”. In: *CVPR*.
- Lee, K.-H., J. Li, A. Gaidon, and G. Ros. (2019c). “SPIGAN: Privileged Adversarial Learning from Simulation”. In: *ICLR*.
- Lee, K., H. Lee, and J. Y. Hwang. (2021). “Self-Mutating Network for Domain Adaptive Segmentation in Aerial Images”. In: *ICCV*.
- Lee, S., H. Seong, S. Lee, and E. Kim. (2022). “WildNet: Learning Domain Generalized Semantic Segmentation from the Wild”. In: *CVPR*.

- Leibe, B., A. Leonardis, and B. Schiel. (2004). “Combined Object Categorization and Segmentation with an Implicit Shape Model”. In: *ECCV Workshop on Statistical Learning in Computer Vision (SLCV)*.
- Lempitsky, V., A. Vedaldi, and A. Zisserman. (2011). “A Pylon Model for Semantic Segmentation”. In: *NeurIPS*.
- Lempitsky, V. S., P. Kohli, C. Rother, and T. Sharp. (2009). “Image Segmentation with A Bounding Box Prior”. In: *ICCV*.
- Lengyel, A., S. Garg, M. Milford, and J. C. van Gemert. (2021). “Zero-Shot Day-Night Domain Adaptation with a Physics Prior”. In: *ICCV*.
- Li, B., Y. Shi, Z. Qi, and Z. Chen. (2018a). “A Survey on Semantic Segmentation”. In: *ICDM Workshops*.
- Li, B., K. Q. Weinberger, S. Belongie, V. Koltun, and R. Ranftl. (2022). “Language-driven Semantic Segmentation”. In: *ICLR*.
- Li, G., G. Kang, W. Liu, Y. Wei, and Y. Yang. (2020a). “Content-Consistent Matching for Domain Adaptive Semantic Segmentation”. In: *ECCV*.
- Li, G., G. Kang, Y. Zhu, Y. Wei, and Y. Yang. (2021a). “Domain Consensus Clustering for Universal Domain Adaptation”. In: *CVPR*.
- Li, H., P. Xiong, J. An, and L. Wang. (2018b). “Pyramid Attention Network for Semantic Segmentation”. In: *BMVC*.
- Li, H., S. Jialin Pan, S. Wang, and A. C. Kot. (2018c). “Domain Generalization with Adversarial Feature Learning”. In: *CVPR*.
- Li, H., T. Löhner, A. Sekuboyina, J. Zhang, B. Wiestler, and B. Menze. (2020b). “Domain Adaptive Medical Image Segmentation via Adversarial Learning of Disease-Specific Spatial Patterns”. arXiv:2001.09313.
- Li, L.-J., R. Socher, and L. Fei-Fei. (2009). “Towards Total Scene Understanding: Classification, Annotation and Segmentation in an Automatic Framework”. In: *CVPR*.
- Li, P., X. Liang, D. Jia, and E. P. Xing. (2018d). “Semantic-aware Grad-GAN for Virtual-to-Real Urban Scene Adaption”. In: *BMVC*.
- Li, Q., A. Arnab, and P. H. S. Torr. (2018e). “Weakly- and Semi-Supervised Panoptic Segmentation”. In: *ECCV*.

- Li, R., Q. Jiao, W. Cao, H.-S. Wong, and S. Wu. (2020c). “Model Adaptation: Unsupervised Domain Adaptation Without Source Data”. In: *CVPR*.
- Li, R., K. Li, Y.-C. Kuo, M. Shu, X. Qi, X. Shen, and J. Jia. (2018f). “Referring Image Segmentation via Recurrent Refinement Networks”. In: *CVPR*.
- Li, S., X. Sui, J. Fu, H. Fu, X. Luo, Y. Feng, X. Xu, Y. Liu, D. Ting, R. Siow, and M. Goh. (2021b). “Few-Shot Domain Adaptation with Polymorphic Transformers”. In: *MICCAI*.
- Li, Y., X. Chen, Z. Zhu, L. Xie, G. Huang, D. Du, and X. Wang. (2019a). “Attention-Guided Unified Network for Panoptic Segmentation”. In: *CVPR*.
- Li, Y., H. Qi, J. Dai, X. Ji, and Y. Wei. (2017). “Fully Convolutional Instance-aware Semantic Segmentation”. In: *CVPR*.
- Li, Y., M. Murias, S. Major, G. Dawson, and D. E. Carlson. (2018g). “Extracting Relationships by Multi-Domain Matching”. In: *NeurIPS*.
- Li, Y., Y. Yang, W. Zhou, and T. M. Hospedales. (2019b). “Feature-Critic Networks for Heterogeneous Domain Generalization”. In: *ICML*.
- Li, Y., L. Yuan, and N. Vasconcelos. (2019c). “Bidirectional Learning for Domain Adaptation of Semantic Segmentation”. In: *CVPR*.
- Li, Z., Z. Chen, F. Yang, W. Li, Y. Zhu, C. Zhao, R. Deng, L. Wu, R. Zhao, M. Tang, and J. Wang. (2021c). “MST: Masked Self-Supervised Transformer for Visual Representation”. In: *NeurIPS*.
- Li, Z., Y. Gan, X. Liang, Y. Yu, H. Cheng, and L. Lin. (2016). “LSTM-CF: Unifying Context Modeling and Fusion with LSTMs for RGB-D Scene Labeling”. In: *ECCV*.
- Lian, Q., F. Lv, L. Duan, and B. Gong. (2019). “Constructing Self-motivated Pyramid Curriculum for Cross-Domain Semantic Segmentation: A Non-Adversarial Approach”. In: *ICCV*.
- Liang, J., D. Hu, and J. Feng. (2020). “Do We Really Need to Access the Source Data? Source Hypothesis Transfer for Unsupervised Domain Adaptation”. In: *ICML*.
- Liang, S., Y. Li, and R. Srikant. (2018a). “Enhancing The Reliability of Out-of-distribution Image Detection in Neural Networks”. In: *ICLR*.

- Liang, X., X. Shen, J. Feng, L. Lin, and S. Yan. (2016). “Semantic Object Parsing with Graph LSTM”. In: *ECCV*.
- Liang, X., Y. Wei, X. Shen, J. Yang, L. Lin, and S. Yan. (2018b). “Proposal-free Network for Instance-level Object Segmentation”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 40(12): 2978–2991.
- Liang, Y., R. Wakaki, S. Nobuhara, and K. Nishino. (2022). “Multimodal Material Segmentation”. In: *CVPR*.
- Lin, D., J. Dai, J. Jia, K. He, and J. Sun. (2016). “ScribbleSup: Scribble-Supervised Convolutional Networks for Semantic Segmentation”. In: *CVPR*.
- Lin, G., A. Milan, C. Shen, and I. Reid. (2017a). “RefineNet: Multi-Path Refinement Networks for High-Resolution Semantic Segmentation”. In: *CVPR*.
- Lin, T.-Y., P. Dollár, R. Girshick, K. He, B. Hariharan, and S. Belongie. (2017b). “Feature Pyramid Networks for Object Detection”. In: *CVPR*.
- Lin, T.-Y., M. Maire, S. Belongie, J. Hays, P. Perona, D. Ramanan, P. Dollár, and C. L. Zitnick. (2014). “Microsoft COCO: Common Objects in Context”. In: *ECCV*.
- Liu, C., J. Yuen, and A. Torralba. (2009). “Nonparametric Scene Parsing: Label Transfer via Dense Scene Alignment”. In: *CVPR*.
- Liu, C., Z. Lin, X. Shen, J. Yang, X. Lu, and A. Yuille. (2017). “Recurrent Multimodal Interaction for Referring Image Segmentation”. In: *ICCV*.
- Liu, F., Z. Zhou, H. Jang, A. Samsonov, G. Zhao, and R. Kijowski. (2018a). “Deep Convolutional Neural Network and 3D Deformable Approach for Tissue Segmentation in Musculoskeletal Magnetic Resonance Imaging”. *Magnetic Resonance in Medicine*. 79: 2379–2391.
- Liu, H., C. Peng, C. Yu, J. Wang, X. Liu, G. Yu, and W. Jiang. (2019). “An End-to-End Network for Panoptic Segmentation”. In: *CVPR*.
- Liu, M.-Y. and O. Tuzel. (2016). “Coupled Generative Adversarial Networks”. In: *NeurIPS*.

- Liu, Q., Q. Dou, L. Yu, and P. A. Heng. (2020a). “MS-Net: Multi-Site Network for Improving Prostate Segmentation with Heterogeneous MRI Data”. *IEEE Transactions on Medical Imaging*. 39(9): 2713–2724.
- Liu, S., L. Qi, H. Qin, J. Shi, and J. Jia. (2018b). “Path Aggregation Network for Instance Segmentation”. In: *CVPR*.
- Liu, W., A. Rabinovich, and A. C. Berg. (2016). “ParseNet: Looking Wider to See Better”. In: *ICLR Workshops*.
- Liu, X., L. Song, S. Liu, and Y. Zhang. (2021a). “A Review of Deep-Learning-Based Medical Image Segmentation Methods”. *Sustainability*. 13(3): 1224.
- Liu, X., Z. Guo, S. Li, F. Xing, J. You, C.-C. J. Kuo, G. El Fakhri, and J. Woo. (2021b). “Adversarial Unsupervised Domain Adaptation With Conditional and Label Shift: Infer, Align and Iterate”. In: *ICCV*.
- Liu, Y., W. Zhang, and J. Wang. (2021c). “Source-Free Domain Adaptation for Semantic Segmentation”. In: *CVPR*.
- Liu, Z., Y. Lin, Y. Cao, H. Hu, Y. Wei, Z. Zhang, S. Lin, and B. Guo. (2021d). “Swin Transformer: Hierarchical Vision Transformer using Shifted Windows”. In: *ICCV*.
- Liu, Z., H. Mao, C.-Y. Wu, C. Feichtenhofer, T. Darrell, and S. Xie. (2022). “A ConvNet for the 2020s”. In: *CVPR*.
- Liu, Z., Z. Miao, X. Pan, X. Zhan, D. Lin, S. X. Yu, and B. Gong. (2020b). “Open Compound Domain Adaptation”. In: *CVPR*.
- Long, J., E. Shelhamer, and T. Darrell. (2015a). “Fully Convolutional Networks for Semantic Segmentation”. In: *CVPR*.
- Long, M., Y. Cao, J. Wang, and M. I. Jordan. (2015b). “Learning Transferable Features with Deep Adaptation Networks”. In: *ICML*.
- Long, M., Z. Cao, J. Wang, and M. I. Jordan. (2018). “Conditional Adversarial Domain Adaptation”. In: *NeurIPS*.
- Lowe, D. G. (2004). “Distinctive Image Features from Scale-invariant Keypoints”. *International Journal of Computer Vision (IJCV)*. 60: 91–110.
- Lucchi, A., Y. Li, X. Boix, K. Smith, and P. Fua. (2011). “Are Spatial and Global Constraints Really Necessary for Segmentation?” In: *ICCV*.

- Luo, G., Y. Zhou, X. Sun, L. Cao, C. Wu, C. Deng, and R. Ji. (2020). “Multi-task Collaborative Network for Joint Referring Expression Comprehension and Segmentation”. In: *CVPR*.
- Luo, W. and M. Yang. (2020). “Semi-supervised Semantic Segmentation via Strong-weak Dual-branch Network”. In: *ECCV*.
- Luo, Y., P. Liu, T. Guan, J. Yu, and Y. Yang. (2019a). “Significance-aware Information Bottleneck for Domain Adaptive Semantic Segmentation”. In: *ICCV*.
- Luo, Y., L. Zheng, T. Guan, J. Yu, and Y. Yang. (2019b). “Taking A Closer Look at Domain Shift: Category-level Adversaries for Semantics Consistent Domain Adaptation”. In: *CVPR*.
- Lv, F., T. Liang, X. Chen, and L. Guosheng. (2020). “Cross-Domain Semantic Segmentation via Domain-Invariant Interactive Relation Transfer”. In: *CVPR*.
- Ma, L., Y. Liu, X. Zhang, Y. Yed, G. Yin, and B. A. Johnson. (2019). “Deep Learning in Remote Sensing Applications: A Meta-analysis and Review”. *ISPRS Journal of Photogrammetry and Remote Sensing*. 152(6): 166–177.
- Ma, X., J. Gao, and C. Xu. (2021). “Active Universal Domain Adaptation”. In: *ICCV*.
- Mansour, Y., M. Mohri, and A. Rostamizadeh. (2009). “Domain Adaptation with Multiple Sources”. In: *NeurIPS*.
- Mao, X., Q. Li, H. Xie, R. Y. K. Lau, Z. Wang, and S. P. Smolley. (2017). “Least Squares Generative Adversarial Networks”. In: *ICCV*.
- Maracani, A., U. Michieli, M. Toldo, and P. Zanuttigh. (2021). “RECALL: Replay-Based Continual Learning in Semantic Segmentation”. In: *ICCV*.
- Martin, D. R., C. C. Fowlkes, and J. Malik. (2004). “Learning to Detect Natural Image Boundaries using Local Brightness, Color, and Texture Cues”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 26(5): 530–549.
- Mei, K., C. Zhu, J. Zou, and S. Zhang. (2020). “Instance Adaptive Self-Training for Unsupervised Domain Adaptation”. arXiv:2008.12197.
- Metzen, J., M. C. Kumar, T. Brox, and V. Fischer. (2017). “Universal Adversarial Perturbations Against Semantic Image Segmentation”. In: *ICCV*.

- Michieli, U. and P. Zanuttigh. (2021). “Knowledge Distillation for Incremental Learning in Semantic Segmentation”. *Computer Vision and Image Understanding (CVIU)*. 205.
- Minaee, S., Y. Boykov, F. Porikli, A. Plaza, N. Kehtarnavaz, and D. Terzopoulos. (2020). “Image Segmentation Using Deep Learning: A Survey”. arXiv:2001.05566.
- Mirikharaji, Z. and G. Hamarneh. (2018). “Star Shape Prior in Fully Convolutional Networks for Skin Lesion Segmentation”. In: *MICCAI*.
- Mittal, S., M. Tatarchenko, and T. Brox. (2021). “Semi-Supervised Semantic Segmentation With High- and Low-Level Consistency”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 43(4): 1369–1379.
- Moeskops, P., M. Veta, M. W. Lafarge, K. A. J. Eppenhof, and J. P. W. Pluim. (2017). “Adversarial Training and Dilated Convolutions for Brain MRI Segmentation”. In: *MICCAI Workshops*.
- Mohan, R. and A. Valada. (2021). “EfficientPS: Efficient Panoptic Segmentation”. *International Journal of Computer Vision (IJCV)*. 129: 1551–1579.
- Moosavi-Dezfooli, S.-M., A. Fawzi, O. Fawzi, and P. Frossard. (2017). “Universal Adversarial Perturbations”. In: *CVPR*.
- Mordan, T., A. Saporta, A. Alahi, M. Cord, and P. Pérez. (2020). “Bilinear Multimodal Discriminator for Adversarial Domain Adaptation with Privileged Information”. In: *Symposium of the European Association for Research in Transportation (hEART)*.
- Mostajabi, M., P. Yadollahpour, and G. Shakhnarovich. (2015). “Feedforward Semantic Segmentation with Zoom-out Features”. In: *CVPR*.
- Motiian, S., M. Piccirilli, D. A. Adjeroh, and G. Doretto. (2017). “Unified Deep Supervised Domain Adaptation and Generalization”. In: *ICCV*.
- Mottaghi, R., X. Chen, X. Liu, N.-G. Cho, S.-W. Lee, S. Fidler, R. Urtasun, and A. Yuille. (2014). “The Role of Context for Object Detection and Semantic Segmentation in the Wild”. In: *CVPR*.
- Mousavian, A., H. Pirsaviash, and J. Košecák. (2016). “Joint Semantic Segmentation and Depth Estimation with Deep Convolutional Networks”. In: *3DV*.

- Murez, Z., S. Kolouri, D. Kriegman, R. Ramamoorthi, and K. Kim. (2018). “Image to Image Translation for Domain Adaptation”. In: *CVPR*.
- Musto, L. and A. Zinelli. (2020). “Semantically Adaptive Image-to-image Translation for Domain Adaptation of Semantic Segmentation”. In: *BMVC*.
- Myronenko, A. (2017). “3D MRI Brain Tumor Segmentation using Autoencoder Regularization”. In: *MICCAI Workshops*.
- Neuhold, G., T. Ollmann, S. R. Bulò, and P. Kotschieder. (2017). “The Mapillary Vistas Dataset for Semantic Understanding of Street Scenes”. In: *ICCV*.
- Nguyen, V.-A., T. Nguyen, T. Le, Q. H. Tran, and D. Phung. (2021). “STEM: An Approach to Multi-Source Domain Adaptation With Guarantees”. In: *ICCV*.
- Nguyen-Meidine, L. T., A. Belal, M. Kiran, J. Dolz, L.-A. Blais-Morin, and E. Granger. (2021). “Unsupervised Multi-Target Domain Adaptation Through Knowledge Distillation”. In: *WACV*.
- Nie, D., L. Wang, E. Adeli, C. Lao, W. Lin, and D. Shen. (2019). “3-D Fully Convolutional Networks for Multimodal Isointense Infant Brain Image Segmentation”. *IEEE Transactions on Computers*. 49(3): 1123–1136.
- Ning, M., D. Lu, D. Wei, C. Bian, C. Yuan, S. Yu, K. Ma, and Y. Zheng. (2021). “Multi-Anchor Active Domain Adaptation for Semantic Segmentation”. In: *ICCV*.
- Noh, H., S. Hong, and B. Han. (2015). “Learning Deconvolution Network for Semantic Segmentation”. In: *ICCV*.
- El-Nouby, A., G. Izacard, H. Touvron, I. Laptev, H. Jegou, and E. Grave. (2021). “Are Large-scale Datasets Necessary for Self-Supervised Pre-training?” arXiv:2112.10740.
- Novikov, A. A., D. Lenis, D. Major, J. Hladuvka, M. Wimmer, and K. Bühler. (2018). “Fully Convolutional Architectures for Multi-Class Segmentation in Chest Radiographs”. *IEEE Transactions on Medical Imaging*. 37: 1865–1876.
- Novosad, P., V. Fonov, and D. L. Collins. (2019). “Unsupervised Domain Adaptation for the Automated Segmentation of Neuroanatomy in MRI: a Deep Learning Approach”. bioRxiv:845537.

- Novotny, D., S. Albanie, D. Larlus, and A. Vedaldi. (2018). “Semi-convolutional Operators for Instance Segmentation”. In: *ECCV*.
- Oktay, O., J. Schlemper, L. Le Folgoc, M. Lee, M. Heinrich, K. Misawa, K. Mori, S. McDonagh, N. Y. Hammerla, B. Kainz, B. Glocker, and D. Rueckert. (2018). “Attention U-Net: Learning Where to Look for the Pancreas”. arXiv:1804.0399.
- Olsson, V., W. Tranheden, J. Pinto, and L. Svensson. (2021). “Class-Mix: Segmentation-based Data Sugmentation for Semi-supervised Learning”. In: *WACV*.
- Oquab, M., L. Bottou, I. Laptev, and J. Sivic. (2015). “Is Object Localization for Free?-weakly-supervised Learning with Convolutional Neural Networks”. In: *CVPR*.
- Orbes-Arteaga, M., T. Varsavsky, C. H. Sudre, Z. Eaton-Rosen, L. Haddow Lewis J. Sorensen, M. Nielsen, A. Pai, S. Ourselin, M. Modat, P. Nachev, and M. J. Cardoso. (2019). “Multi-Domain Adaptation in Brain MRI through Paired Consistency and Adversarial Learning”. In: *MICCAI Workshops*.
- Orbanz, P. and J. M. Buhmann. (2006). “Smooth Image Segmentation by Nonparametric Bayesian Inference”. In: *ECCV*.
- Ouali, Y., C. Hudelot, and M. Tami. (2020). “Semi-supervised Semantic Segmentation with Cross-consistency Training”. In: *CVPR*.
- Pan, F., I. Shin, F. Rameau, S. Lee, and I. Kweon. (2020). “Unsupervised Intra-domain Adaptation for Semantic Segmentation through Self-Supervision”. In: *CVPR*.
- Panareda Busto, P. and J. Gall. (2017). “Open Set Domain Adaptation”. In: *ICCV*.
- Pantofaru, C., C. Schmid, and M. He. (2008). “Object Recognition by Integrating Multiple Image Segmentations”. In: *ECCV*.
- Papandreou, G., L.-C. Chen, K. P. Murphy, and A. L. Yuille. (2015). “Weakly-and Semi-Supervised Learning of a Deep Convolutional Network for Semantic Image Segmentation”. In: *ICCV*.
- Park, T., M.-Y. Liu, T.-C. Wang, and J.-Y. Zhu. (2019). “Semantic Image Synthesis with Spatially-Adaptive Normalization”. In: *CVPR*.
- Paszke, A., A. Chaurasia, S. Kim, and E. Culurciello. (2016). “ENet: A Deep Neural Network Architecture for Real-Time Semantic Segmentation”. arXiv:1606.02147.

- Pathak, D., P. Krüahenbühl, and T. Darrell. (2015). “Constrained Convolutional Neural Networks for Weakly Supervised Segmentation”. In: *ICCV*.
- Peng, X., Q. Bai, X. Xia, Z. Huang, K. Saenko, and B. Wang. (2019). “Moment Matching for Multi-Source Domain Adaptation”. In: *ICCV*.
- Peng, X., Y. Li, and K. Saenko. (2020). “Domain2Vec: Domain Embedding for Unsupervised Domain Adaptation”. In: *ECCV*.
- Perone, C. S., P. Ballester, R. C. Barros, and J. Cohen-Adad. (2019). “Unsupervised Domain Adaptation for Medical Imaging Segmentation with Self-Ensembling”. *NeuroImage*. 194: 1–11.
- Perronnin, F. and C. R. Dance. (2007). “Fisher Kernels on Visual Vocabularies for Image Categorization”. In: *CVPR*.
- Pinheiro, P. H. O. and R. Collobert. (2014). “Recurrent Convolutional Neural Networks for Scene Parsing”. In: *ICML*.
- Pinheiro, P. O., R. Collobert, and P. Dollár. (2015). “Learning to Segment Object Candidates”. In: *NeurIPS*.
- Plath, N., M. Toussaint, and S. Nakajima. (2009). “Multi-class Image Segmentation using Conditional Random Fields and Global Classification”. In: *ICML*.
- Pohlen, T., A. Hermans, M. Mathias, and B. Leibe. (2017). “Full-Resolution Residual Networks for Semantic Segmentation in Street Scenes”. In: *CVPR*.
- Pont-Tuset, J., P. Arbeláez, J. T. Barron, F. Marques, and J. Malik. (2016). “Multiscale Combinatorial Grouping for Image Segmentation and Object Proposal Generation”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 39(1): 128–140.
- Porav, H., T. Bruls, and P. Newman. (2019). “Don’t Worry About the Weather: Unsupervised Condition-Dependent Domain Adaptation”. In: *IEEE Intelligent Transportation Systems Conference*.
- Porzi, L., S. R. Buló, A. Colovic, and P. Kotschieder. (2019). “Seamless Scene Segmentation”. In: *CVPR*.
- Qi, X., R. Liao, J. Jia, S. Fidler, and R. Urtasun. (2017). “3D Graph Neural Networks for RGBD Semantic Segmentation”. In: *ICCV*.
- Qi, X., Z. Liu, J. Shi, H. Zhao, and J. Jia. (2016). “Augmented Feedback in Semantic Segmentation under Image Level Supervision”. In: *ECCV*.

- Qiao, F., L. Zhao, and X. Peng. (2020). “Learning to Learn Single Domain Generalization”. In: *CVPR*.
- Radford, A., J. W. Kim, C. Hallacy, A. Ramesh, G. Goh, S. Agarwal, A. Sastry Girishand Askell, P. Mishkin, J. Clark, G. Krueger, and I. Sutskever. (2021). “Learning Transferable Visual Models From Natural Language Supervision”. In: *ICML*.
- Rahman, M. A. and Y. Wang. (2017). “Optimizing Intersection-Over-Union in Deep Neural Networks for Image Segmentation”. In: *International Symposium on Visual Computing*.
- Rahman, M. M., C. Fookes, M. Baktashmotlagh, and S. Sridharan. (2020). “Correlation-aware Adversarial Domain Adaptation and Generalization”. *Pattern Recognition*. 100(107124).
- Rakshit, S., D. Tamboli, P. S. Meshram, B. Banerjee, G. Roig, and S. Chaudhuri. (2020). “Multi-source Open-Set Deep Adversarial Domain Adaptation”. In: *ECCV*.
- Ranftl, R., A. Bochkovskiy, and V. Koltun. (2021). “Vision Transformers for Dense Prediction”. In: *ICCV*.
- Redondo-Cabrera, C., M. Baptista-Ríos, and R. López-Sastre. (2018). “Learning to Exploit the Prior Network Knowledge for Weakly-Supervised Semantic Segmentation”. *IEEE Transactions on Image Processing (TIP)*. 28: 3649–3661.
- Ren, M. and R. S. Zemel. (2017). “End-to-End Instance Segmentation with Recurrent Attention”. In: *CVPR*.
- Ren, S., K. He, R. Girshick, and J. Sun. (2015). “Faster R-CNN: Towards Real-time Object Detection with Region Proposal Networks”. In: *NeurIPS*.
- Rezaei, M., K. Harmuth, W. Gierke, T. Kellermeier, M. Fischer, H. Yang, and C. Meinel. (2017). “Conditional Adversarial Network for Semantic Segmentation of Brain Tumor”. In: *MICCAI Workshops*.
- Richter, S. R., V. Vineet, S. Roth, and K. Vladlen. (2016). “Playing for Data: Ground Truth from Computer Games”. In: *ECCV*.
- Ronneberger, O., P. Fischer, and T. Brox. (2015). “U-Net: Convolutional Networks for Biomedical Image Segmentation”. In: *MICCAI*.
- Ros, G., L. Sellart, J. Materzyńska, D. Vázquez, and A. M. López. (2016). “The SYNTHIA Dataset: a Large Collection of Synthetic Images for Semantic Segmentation of Urban Scenes”. In: *CVPR*.

- Roth, P. M., S. Sternig, H. Grabner, and H. Bischof. (2009). “Classifier grids for robust adaptive object detection”. In: *CVPR*.
- Rother, C., V. Kolmogorov, and A. Blake. (2004). “GrabCut — Interactive Foreground Extraction using Iterated Graph Cuts”. *IEEE Transactions on Graphics*. 23(3): 309–314.
- Roy, A. and S. Todorovic. (2017). “Combining Bottom-Up, Top-Down, and Smoothness Cues for Weakly Supervised Image Segmentation”. In: *CVPR*.
- Roy, S., E. Krivosheev, Z. Zhong, N. Sebe, and E. Ricci. (2021). “Curriculum Graph Co-Teaching for Multi-Target Domain Adaptation”. In: *CVPR*.
- Ru, L., Y. Zhan, B. Yu, and B. Du. (2022). “Learning Affinity from Attention: End-to-End Weakly-Supervised Semantic Segmentation with Transformers”. In: *CVPR*.
- Russell, B. C., A. Torralba, K. P. Murphy, and W. T. Freeman. (2008). “LabelMe: a Database and Web-based Tool for Image Annotation”. *International Journal of Computer Vision (IJCV)*. 77: 157–173.
- Russo, P., T. Tommasi, and B. Caputo. (2019). “Towards Multi-source Adaptive Semantic Segmentation”. In: *ICIAP*.
- Saito, K., D. Kim, S. Sclaroff, T. Darrell, and K. Saenko. (2021). “Tune It the Right Way: Unsupervised Validation of Domain Adaptation via Soft Neighborhood Density”. In: *ICCV*.
- Saito, K. and K. Saenko. (2021). “OVANet: One-vs-All Network for Universal Domain Adaptation”. In: *ICCV*.
- Saito, K., Y. Ushiku, T. Harada, and K. Saenko. (2018a). “Adversarial Dropout Regularization”. In: *ICLR*.
- Saito, K., K. Watanabe, Y. Ushiku, and T. Harada. (2018b). “Maximum Classifier Discrepancy for Unsupervised Domain Adaptation”. In: *CVPR*.
- Saito, K., S. Yamamoto, Y. Ushiku, and T. Harada. (2018c). “Open Set Domain Adaptation by Backpropagation”. In: *ECCV*.
- Sakaridis, C., D. Dai, S. Hecker, and L. Van Gool. (2018). “Model Adaptation with Synthetic and Real Data for Semantic Dense Foggy Scene Understanding”. In: *ECCV*.

- Sakaridis, C., D. Dai, and L. Van Gool. (2019). “Guided Curriculum Model Adaptation and Uncertainty-Aware Evaluation for Semantic Nighttime Image Segmentation”. In: *ICCV*.
- Sakaridis, C., D. Dai, and L. Van Gool. (2021). “ACDC: The Adverse Conditions Dataset with Correspondences for Semantic Driving Scene Understanding”. In: *ICCV*.
- Salamati, N., D. Larlus, G. Csurka, and S. Süssstrunk. (2014). “Incorporating Near-Infrared Information into Semantic Image Segmentation”. arXiv:1406.6147.
- Sankaranarayanan, S., Y. Balaji, A. Jain, S. Nam Lim, and R. Chellappa. (2018). “Learning from Synthetic Data: Addressing Domain Shift for Semantic Segmentation”. In: *CVPR*.
- Saporta, A., T.-H. Vu, M. Cord, and P. Pérez. (2020). “ESL: Entropy-guided Self-supervised Learning for Domain Adaptation in Semantic Segmentation”. In: *CVPR Workshops*.
- Saporta, A., T.-H. Vu, M. Cord, and P. Pérez. (2021). “Multi-Target Adversarial Frameworks for Domain Adaptation in Semantic Segmentation”. In: *ICCV*.
- Schneider, L., M. Jasch, B. Fröhlich, T. Weber, U. Franke, M. Pollefeys, and M. Rätzsch. (2017). “Multimodal Neural Networks: RGB-D for Semantic Segmentation and Object Detection”. In: *SCIA*.
- Schneider, S., E. Rusak, L. Eck, O. Bringmann, W. Brendel, and M. Bethge. (2020). “Improving robustness against common corruptions by covariate shift adaptation”. In: *NeurIPS*.
- Schroff, F., A. Criminisi, and A. Zisserman. (2006). “Single-Histogram Class Models for Image Segmentation”. In: *ICCVGIP*.
- Schwing, A. G. and R. Urtasun. (2015). “Fully Connected Deep Structured Networks”. arXiv:1506.04579.
- Selvaraju, R. R., M. Cogswell, A. Das, R. Vedantam, D. Parikh, and D. Batra. (2017). “Grad-CAM: Visual Explanations from Deep Networks via Gradient-based Localization”. In: *ICCV*.
- Seo, S., Y. Suh, D. Kim, and B. Han. (2020). “Learning to Optimize Domain Specific Normalization for Domain Generalization”. In: *ECCV*.

- Shen, T., D. Gong, W. Zhang, C. Shen, and T. Mei. (2019). “Regularizing Proxies with Multi-Adversarial Training for Unsupervised Domain-Adaptive Semantic Segmentation”. arXiv:1907.12282.
- Shen, T., G. Lin, L. Liu, C. Shen, and I. Reid. (2017). “Weakly Supervised Semantic Segmentation Based on Web Image Co-segmentation”. In: *BMVC*.
- Shen, T., G. Lin, C. Shen, and I. Reid. (2018). “Bootstrapping the Performance of Webly Supervised Semantic Segmentation”. In: *CVPR*.
- Shin, I., D.-J. Kim, J. Cho, S. Woo, K. Park, and I. S. Kweon. (2021). “LabOR: Labeling Only if Required for Domain Adaptive Semantic Segmentation”. In: *ICCV*.
- Shin, I., S. Woo, F. Pan, and I. S. Kweon. (2020). “Two-Phase Pseudo Label Densification for Self-training Based Domain Adaptation”. In: *ECCV*.
- Shotton, J., J. Winn, C. Rother, and A. Criminisi. (2009). “Texton-Boost for Image Understanding: Multi-Class Object Recognition and Segmentation by Jointly Modeling Texture, Layout, and Context”. *International Journal of Computer Vision (IJCV)*. 81: 2–23.
- Silberman, N., D. Hoiem, P. Kohli, and R. Fergus. (2012). “Indoor Segmentation and Support Inference from RGBD Images”. In: *ECCV*.
- Singh, R., M. Vatsa, V. M. Patel, and N. Ratha, eds. (2020). *Domain Adaptation for Visual Understanding. Image Processing, Computer Vision, Pattern Recognition & Graphics*. Springer.
- Sivaprasad, P. T. and F. Fleuret. (2021). “Uncertainty Reduction for Model Adaptation in Semantic Segmentation”. In: *CVPR*.
- Sofiuk, K., O. Barinova, and A. Konushin. (2019). “AdaptIS: Adaptive Instance Selection Network”. In: *ICCV*.
- Song, C., Y. Huang, W. Ouyang, and L. Wang. (2019). “Box-driven Class-wise Region Masking and Filling Rate Guided Loss for Weakly Supervised Semantic Segmentation”. In: *CVPR*.
- Souly, N., C. Spampinato, and M. Shah. (2017). “Semi-supervised Semantic Segmentation using Generative Adversarial Network”. In: *ICCV*.
- Soviany, P., R. Ionescu, P. Rota, and N. Sebe. (2021). “Curriculum Learning: A Survey”. arXiv:2101.10382.

- Strudel, R., R. Garcia, I. Laptev, and C. Schmid. (2021). “Segmenter: Transformer for Semantic Segmentation”. In: *ICCV*.
- Sun, G., W. Wang, J. Dai, and L. Van Gool. (2020a). “Mining Cross-Image Semantics for Weakly Supervised Semantic Segmentation”. In: *ECCV*.
- Sun, K., B. Xiao, D. Liu, and J. Wang. (2019a). “Deep High-Resolution Representation Learning for Human Pose Estimation”. In: *CVPR*.
- Sun, K., Y. Zhao, B. Jiang, T. Cheng, B. Xiao, D. Liu, Y. Mu, X. Wang, W. Liu, and J. Wang. (2019b). “High-Resolution Representations for Labeling Pixels and Regions”. arXiv:1904.04514.
- Sun, Y., E. Tzeng, T. Darrell, and A. A. Efros. (2019c). “Unsupervised Domain Adaptation through Self-Supervision”. arXiv:1909.11825.
- Sun, Y., X. Wang, Z. Liu, J. Miller, A. Efros, and M. Hardt. (2020b). “Test-Time Training with Self-Supervision for Generalization under Distribution Shifts”. In: *ICML*.
- Sun, Y., W. Zuo, and M. Liu. (2019d). “RTFNet: RGB-Thermal Fusion Network for Semantic Segmentation of Urban Scenes”. *IEEE Robotics and Automation Letters*. 4(3): 2576–2583.
- Szegedy, C., W. Zaremba, I. Sutskever, J. Bruna, D. Erhan, I. Goodfellow, and R. Fergus. (2014). “Intriguing Properties of Neural Networks”. In: *ICLR*.
- Taigman, Y., A. Polyak, and L. Wolf. (2017). “Unsupervised Cross-domain Image Generation”. In: *ICLR*.
- Tang, M., F. Perazzi, A. Djelouah, I. Ben Ayed, C. Schroers, and Y. Boykov. (2018). “On Regularized Losses for Weakly-supervised CNN Segmentation”. In: *ECCV*.
- Tarvainen, A. and H. Valpola. (2017). “Mean Teachers are Better Role Models: Weight-averaged Consistency Targets Improve Semi-supervised Deep Learning Results”. In: *NeurIPS*.
- Tasar, O., Y. Tarabalka, A. Giros, P. Alliez, and S. Clerc. (2020). “StandardGAN: Multi-Source Domain Adaptation for Semantic Segmentation of Very High Resolution Satellite Images by Data Standardization”. In: *CVPR Workshops*.
- Teichmann, M. and R. Cipolla. (2019). “Convolutional CRFs for Semantic Segmentation”. In: *BMVC*.

- Teichmann, M., M. Weber, M. Zöllner, R. Cipolla, and R. Urtasun. (2018). “MultiNet: Real-time Joint Semantic Reasoning for Autonomous Driving”. In: *IEEE Intelligent Vehicles Symposium (IVS)*.
- Thoma, M. (2016). “A Survey of Semantic Segmentation”. arXiv:1602.06541.
- Thomas, C. and A. Kovashka. (2019). “Artistic Object Recognition by Unsupervised Style Adaptation”. In: *ACCV*.
- Tian, Z., C. Shen, and H. Chen. (2020). “Conditional Convolutions for Instance Segmentation”. In: *ECCV*.
- Tian, Z., C. Shen, X. Wang, and H. Chen. (2019). “Weakly Supervised Instance Segmentation using the Bounding Box Tightness Prior”. In: *NeurIPS*.
- Tian, Z., C. Shen, X. Wang, and H. Chen. (2021). “BoxInst: High-Performance Instance Segmentation with Box Annotations”. In: *CVPR*.
- Tighe, J., M. Niethammer, and S. Lazebnik. (2014). “Scene Parsing with Object Instances and Occlusion Ordering”. In: *CVPR*.
- Tobin, J., R. Fong, A. Ray, J. Schneider, W. Zaremba, and P. Abbeel. (2017). “Domain Randomization for Transferring Deep Neural Networks from Simulation to the Real World”. In: *IROS*.
- Toldo, M., A. Maracani, U. Michieli, and P. Zanuttigh. (2020a). “Unsupervised Domain Adaptation in Semantic Segmentation: a Review”. arXiv:2005.10876.
- Toldo, M., U. Michieli, G. Agresti, and P. Zanuttigh. (2020b). “Unsupervised Domain Adaptation for Mobile Semantic Segmentation based on Cycle Consistency and Feature Alignment”. *Image and Vision Computing*. 95(103889).
- Toldo, M., U. Michieli, and P. Zanuttigh. (2021). “Unsupervised Domain Adaptation in Semantic Segmentation via Orthogonal and Clustered Embeddings”. In: *WACV*.
- Tran, T., O.-H. Kwon, K.-R. Kwon, S.-H. Lee, and K.-W. Kang. (2018). “Blood Cell Images Segmentation using Deep Learning Semantic Segmentation”. In: *IEEE International Conference on Electronics and Communication Engineering*.
- Tranheden, W., V. Olsson, J. Pinto, and L. Svensson. (2021). “DACS: Domain Adaptation via Cross-domain Mixed Sampling”. In: *WACV*.

- Truong, T.-D., C. N. Duong, N. Le, S. L. Phung, C. Rainwater, and K. Luu. (2021). “BiMaL: Bijective Maximum Likelihood Approach to Domain Adaptation in Semantic Scene Segmentation”. In: *ICCV*.
- Tsai, Y.-H., W.-C. Hung, S. Schulter, K. Sohn, M.-H. Yang, and M. Chandraker. (2018). “Learning to Adapt Structured Output Space for Semantic Segmentation”. In: *CVPR*.
- Tsai, Y.-H., K. Sohn, S. Schulter, and M. Chandraker. (2019). “Domain Adaptation for Structured Output via Discriminative Patch Representations”. In: *ICCV*.
- Turkmen, S. (2019). “Scene Understanding Through Semantic Image Segmentation in Augmented Reality”. *Tech. rep.* University of Oulu.
- Tzeng, E., J. Hoffman, T. Darrell, and K. Saenko. (2015). “Simultaneous Deep Transfer Across Domains and Tasks”. In: *ICCV*.
- Tzeng, E., J. Hoffman, K. Saenko, and T. Darrell. (2017). “Adversarial Discriminative Domain Adaptation”. In: *CVPR*.
- Ulyanov, D., A. Vedaldi, and V. S. Lempitsky. (2016). “Instance Normalization: The Missing Ingredient for Fast Stylization”. arXiv:1607.08022.
- Valada, A., J. Vertens, A. Dhall, and W. Burgard. (2017). “AdapNet: Adaptive Semantic Segmentation in Adverse Environmental Conditions”. In: *ICRA*.
- Valindria, V. V., I. Lavdas, W. Bai, K. Kamnitsas, E. O. Aboagye, A. G. Rockall, D. Rueckert, and B. Glocker. (2018). “Domain Adaptation for MRI Organ Segmentation using Reverse Classification Accuracy”. In: *Medical Imaging with Deep Learning*.
- Varma, G., A. Subramanian, A. Namboodiri, M. Chandraker, and C. Jawahar. (2019). “IDD: A Dataset for Exploring Problems of Autonomous Navigation in Unconstrained Environments”. In: *WACV*.
- Vaswani, A., P. Ramachandran, A. Srinivas, N. Parmar, B. Hechtman, and J. Shlens. (2021). “Scaling Local Self-Attention for Parameter Efficient Visual Backbones”. In: *CVPR*.
- Venkataramani, R., H. Ravishankar, and S. Anamandra. (2019). “Towards Continuous Domain Adaptation For Medical Imaging”. In: *International Symposium on Biomedical Imaging (ISBI)*.
- Venkateswara, H. and S. Panchanathan, eds. (2020). *Domain Adaptation in Computer Vision with Deep Learning*. Springer.

- Verbeek, J. and B. Triggs. (2007a). “Region Classification with Markov Field Aspect Models”. In: *CVPR*.
- Verbeek, J. and B. Triggs. (2007b). “Scene Segmentation with CRFs Learned from Partially Labeled Images”. In: *NeurIPS*.
- Vernaza, P. and M. Chandraker. (2017). “Learning Random-walk Label Propagation for Weakly-supervised Semantic Segmentation”. In: *CVPR*.
- Vezhnevets, A. and J. M. Buhmann. (2010). “Towards Weakly Supervised Semantic Segmentation by Means of Multiple Instance and Multitask Learning”. In: *CVPR*.
- Visin, F., M. Ciccone, A. Romero, K. Kastner, K. Cho, Y. Bengio, M. Matteucci, and A. Courville. (2016). “ReSeg: A Recurrent Neural Network-based Model for Semantic Segmentation”. In: *CVPR Workshops*.
- Volpi, R., P. De Jorge, D. Larlus, and G. Csurka. (2022). “On the Road to Online Adaptation for Semantic Image Segmentation”. In: *CVPR*.
- Volpi, R., D. Larlus, and G. Rogez. (2021). “Continual Adaptation of Visual Representations via Domain Randomization and Meta-Learning”. In: *CVPR*.
- Volpi, R. and V. Murino. (2019). “Model Vulnerability to Distributional Shifts over Image Transformation Sets”. In: *ICCV*.
- Volpi, R., H. Namkoong, O. Sener, J. C. Duchi, V. Murino, and S. Savarese. (2019). “Generalizing to Unseen Domains via Adversarial Data Augmentation”. In: *NeurIPS*.
- Vu, T.-H., H. Jain, M. Bucher, M. Cord, and P. Pérez. (2019a). “ADVENT: Adversarial Entropy Minimization for Domain Adaptation in Semantic Segmentation”. In: *CVPR*.
- Vu, T.-H., H. Jain, M. Bucher, M. Cord, and P. Pérez. (2019b). “DADA: Depth-Aware Domain Adaptation in Semantic Segmentation”. In: *ICCV*.
- Wang, D., E. Shelhamer, S. Liu, B. Olshausen, and T. Darrell. (2021a). “TENT: Fully Test-Time Adaptation by Entropy Minimization”. In: *ICLR*.
- Wang, D., C. Gu, K. Wu, and X. Guan. (2017). “Adversarial Neural Networks for Basal Membrane Segmentation of Microinvasive Cervix Carcinoma in Histopathology Images”. In: *ICMLC*.

- Wang, H., Y. Zhu, B. Green, H. Adam, A. Yuille, and L.-C. Chen. (2020a). “Axial-DeepLab: Stand-Alone Axial-Attention for Panoptic Segmentation”. In: *ECCV*.
- Wang, J., C. Lan, C. Liu, Y. Ouyang, and T. Qin. (2020b). “Generalizing to Unseen Domains: A Survey on Domain Generalization”. arXiv:2103.03097.
- Wang, K., C. Yang, and M. Betke. (2021b). “Consistency Regularization with High-dimensional Non-adversarial Source-guided Perturbation for Unsupervised Domain Adaptation in Segmentation”. In: *AAAI*.
- Wang, M. and W. Deng. (2018). “Deep Visual Domain Adaptation: A Survey”. *Neurocomputing*. 312: 135–153.
- Wang, P., P. Chen, Y. Yuan, D. Liu, Z. Huang, X. Hou, and G. Cottrell. (2018a). “Understanding Convolution for Semantic Segmentation”. In: *WACV*.
- Wang, P., X. Shen, Z. Lin, S. Cohen, B. Price, and A. Yuille. (2015). “Towards Unified Depth and Semantic Prediction from a Single Image”. In: *CVPR*.
- Wang, Q., O. Fink, L. Van Gool, and D. Dai. (2022a). “Continual Test-Time Domain Adaptation”. In: *CVPR*.
- Wang, S., L. Yu, C. Li, C.-W. Fu, and P.-A. Heng. (2020c). “Learning from Extrinsic and Intrinsic Supervisions for Domain Generalization”. In: *ECCV*.
- Wang, W., E. Xie, X. Li, D.-P. Fan, K. Song, D. Liang, T. Lu, P. Luo, and L. Shao. (2021c). “Pyramid Vision Transformer: A Versatile Backbone for Dense Prediction without Convolutions”. In: *ICCV*.
- Wang, X., S. You, X. Li, and H. Ma. (2018b). “Weakly-Supervised Semantic Segmentation by Iteratively Mining Common Object Features”. In: *CVPR*.
- Wang, X., T. Kong, C. Shen, Y. Jiang, and L. Li. (2020d). “SOLO: Segmenting Objects by Locations”. In: *ECCV*.
- Wang, Y., J. Peng, and Z. Zhang. (2021d). “Uncertainty-Aware Pseudo Label Refinery for Domain Adaptive Semantic Segmentation”. In: *ICCV*.
- Wang, Z., Y. Lu, Q. Li, X. Tao, Y. Guo, M. Gong, and T. Liu. (2022b). “CRIS: CLIP-Driven Referring Image Segmentation”. In: *CVPR*.

- Wang, Z., Y. Wei, R. Feris, J. Xiong, W.-M. Hwu, T. S. Huang, and H. Shi. (2020e). “Alleviating Semantic-level Shift: A Semi-supervised Domain Adaptation Method for Semantic Segmentation”. In: *CVPR Workshops*.
- Wang, Z., M. You, Y. Wei, R. Feris, J. Xiong, W.-m. Hwu, T. S. Huang, and H. Shi. (2020f). “Differential Treatment for Stuff and Things: A Simple Unsupervised Domain Adaptation Method for Semantic Segmentation”. In: *CVPR*.
- Wei, C., H. Fan, S. Xie, C. Wu, A. L. Yuille, and C. Feichtenhofer. (2021). “Masked Feature Prediction for Self-Supervised Visual Pre-Training”. arXiv:2112.09133.
- Wilson, G. and D. J. Cook. (2020). “A Survey of Unsupervised Deep Domain Adaptation”. *IEEE Transactions on Intelligent Systems and Technology*. 11(5).
- Winn, J. and J. Shotton. (2006). “The Layout Consistent Random Field for Recognizing and Segmenting Partially Occluded Objects”. In: *CVPR*.
- Wrenninge, M. and J. Unger. (2018). “Synscapes: A Photorealistic Synthetic Dataset for Street Scene Parsing”. arXiv:1810.08705.
- Wu, F.-Y. (1982). “The Potts Model”. *Reviews of Modern Physics*. 54(1): 235–268.
- Wu, Z., C. Shen, and A. van den Hengel. (2017). “Real-time Semantic Image Segmentation via Spatial Sparsity”. arXiv:1712.00213.
- Wu, Z., X. Han, Y.-L. Lin, M. Gokhan Uzunbas, T. Goldstein, S. Nam Lim, and L. S. Davis. (2018). “DCAN: Dual Channel-wise Alignment Networks for Unsupervised Scene Adaptation”. In: *ECCV*.
- Wu, Z., X. Wang, J. E. Gonzalez, T. Goldstein, and L. S. Davis. (2019). “ACE: Adapting to Changing Environments for Semantic Segmentation”. In: *ICCV*.
- Wulfmeier, M., A. Bewley, and I. Posner. (2017). “Addressing Appearance Change in Outdoor Robotics with Adversarial Domain Adaptations”. In: *IROS*.
- Xia, H., H. Zhao, and Z. Ding. (2021). “Adaptive Adversarial Network for Source-Free Domain Adaptation”. In: *ICCV*.
- Xia, W., C. Domokos, J. Dong, L.-F. Cheong, and S. Yan. (2013). “Semantic Segmentation without Annotating Segments”. In: *CVPR*.

- Xiao, T., Y. Liu, B. Zhou, Y. Jiang, and J. Sun. (2018). “Unified Perceptual Parsing for Scene Understanding”. In: *ECCV*.
- Xie, B., L. Yuan, S. Li, C. H. Liu, and X. Cheng. (2022a). “Towards Fewer Annotations: Active Learning via Region Impurity and Prediction Uncertainty for Domain Adaptive Semantic Segmentation”. In: *CVPR*.
- Xie, C., J. Wang, Z. Zhang, Y. Zhou, L. Xie, and A. L. Yuille. (2017). “Adversarial Examples for Semantic Segmentation and Object Detection”. In: *ICCV*.
- Xie, E., W. Wang, Z. Yu, A. Anandkumar, J. M. Alvarez, and P. Luo. (2021). “SegFormer: Simple and Efficient Design for Semantic Segmentation with Transformers”. In: *NeurIPS*.
- Xie, Z., Z. Zhang, Y. Cao, Y. Lin, J. Bao, Z. Yao, Q. Dai, and H. Hu. (2022b). “SimMIM: A Simple Framework for Masked Image Modeling”. In: *CVPR*.
- Xiong, Y., R. Liao, H. Zhao, R. Hu, M. Bai, E. Yumer, and R. Urtasun. (2019). “UPSNet: A Unified Panoptic Segmentation Network”. In: *CVPR*.
- Xu, D., W. Ouyang, X. Wang, and N. Sebe. (2018). “PAD-Net: Multi-Tasks Guided Prediction-and-Distillation Network for Simultaneous Depth Estimation and Scene Parsing”. In: *CVPR*.
- Xu, J., A. G. Schwing, and R. Urtasun. (2015). “Learning to Segment Under Various Forms of Weak Supervision”. In: *CVPR*.
- Xu, J., L. Xiao, and A. M. López. (2019a). “Self-Supervised Domain Adaptation for Computer Vision Tasks”. *IEEE Access*. 7: 156694–156706.
- Xu, L., W. Ouyang, M. Bennamoun, F. Boussaid, and D. Xu. (2022). “Multi-class Token Transformer for Weakly Supervised Semantic Segmentation”. In: *CVPR*.
- Xu, W., Y. Xu, T. Chang, and Z. Tu. (2021). “Co-Scale Conv-Attentional Image Transformers”. In: *ICCV*.
- Xu, Y., B. Du, L. Zhang, Q. Zhang, G. Wang, and L. Zhang. (2019b). “Self-Ensembling Attention Networks: Addressing Domain Shift for Semantic Segmentation”. In: *AAAI*.
- Xu, Z., W. Li, L. Niu, and D. Xu. (2014). “Exploiting Low-rank Structure from Latent Domains for Domain Generalization”. In: *ECCV*.

- Yang, J., R. Xu, R. Li, X. Qi, X. Shen, G. Li, and L. Lin. (2020a). “An Adversarial Perturbation Oriented Domain Adaptation Approach for Semantic Segmentation”. In: *AAAI*.
- Yang, J., W. An, S. Wang, X. Zhu, C. Yan, and J. Huang. (2020b). “Label-driven Reconstruction for Domain Adaptation in Semantic Segmentation”. In: *ECCV*.
- Yang, J., C. Li, W. An, H. Ma, Y. Guo, Y. Rong, P. Zhao, and J. Huang. (2021). “Exploring Robustness of Unsupervised Domain Adaptation in Semantic Segmentation”. In: *ICCV*.
- Yang, L., W. Zhuo, ei Qi, Y. Shi, and Y. Gao. (2022). “ST++: Make Self-training Work Better for Semi-supervised Semantic Segmentation”. In: *CVPR*.
- Yang, L., P. Meer, and D. J. Foran. (2007). “Multiple Class Segmentation Using A Unified Framework over Mean-Shift Patches”. In: *CVPR*.
- Yang, L., Y. Balaji, S.-N. Lim, and A. Shrivastava. (2020c). “Curriculum Manager for Source Selection in Multi-Source Domain Adaptation”. In: *ECCV*.
- Yang, M., K. Yu, C. Zhang, Z. Li, and K. Yang. (2018). “DenseASPP for Semantic Segmentation in Street Scenes”. In: *CVPR*.
- Yang, T.-J., M. D. Collins, Y. Zhu, J.-J. Hwang, T. Liu, X. Zhang, V. Sze, G. Papandreou, and L.-C. Chen. (2019). “DeeperLab: Single-Shot Image Parser”. arXiv:1902.05093.
- Yang, Y., D. Lao, G. Sundaramoorthi, and S. Soatto. (2020d). “Phase Consistent Ecological Domain Adaptation”. In: *CVPR*.
- Yang, Y. and S. Soatto. (2020). “FDA: Fourier Domain Adaptation for Semantic Segmentation”. In: *CVPR*.
- Yang, Y., S. Hallman, D. Ramanan, and C. C. Fowlkes. (2012). “Layered Object Models for Image Segmentation”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 34(9): 1731–1743.
- Ye, C., W. Wang, S. Zhang, and K. Wang. (2019a). “Multi-Depth Fusion Network for Whole-Heart CT Image Segmentation”. *IEEE Access*. 7: 23421–23429.
- Ye, L., M. Rochan, Z. Liu, and Y. Wang. (2019b). “Cross-Modal Self-Attention Network for Referring Image Segmentation”. In: *CVPR*.

- You, K., X. Wang, M. Long, and M. I. Jordan. (2019). “Towards Accurate Model Selection in Deep Unsupervised Domain Adaptation”. In: *ICML*.
- Yu, C., J. Wang, C. Peng, C. Gao, G. Yu, and N. Sang. (2018a). “BiSeNet: Bilateral Segmentation Network for Real-time Semantic Segmentation”. In: *ECCV*.
- Yu, F., H. Chen, X. Wang, W. Xian, Y. Chen, F. Liu, V. Madhavan, and T. Darrell. (2020). “BDD100K: A Diverse Driving Dataset for Heterogeneous Multitask Learning”. In: *CVPR*.
- Yu, F. and V. Koltun. (2015). “Multi-Scale Context Aggregation by Dilated Convolutions”. arXiv:1511.07122.
- Yu, L., Z. Lin, X. Shen, J. Yang, X. Lu, M. Bansal, and T. L. Berg. (2018b). “MAttNet: Modular Attention Network for Referring Expression Comprehension”. In: *CVPR*.
- Yuan, J., Y. Liu, C. Shen, Z. Wang, and H. Li. (2021a). “A Simple Baseline for Semi-Supervised Semantic Segmentation With Strong Data Augmentation”. In: *ICCV*.
- Yuan, L., D. Chen, Y.-L. Chen, N. Codella, X. Dai, J. Gao, H. Hu, X. Huang, B. Li, C. Li, C. Liu, M. Liu, Z. Liu, Y. Lu, Y. Shi, L. Wang, J. Wang, B. Xiao, Z. Xiao, J. Yang, M. Zeng, L. Zhou, and P. Zhang. (2021b). “Florence: A New Foundation Model for Computer Vision”. arXiv:2111.11432.
- Yuan, Y., X. Chen, and J. Wang. (2020). “Object-Contextual Representations for Semantic Segmentation”. In: *ECCV*.
- Yue, X., Y. Zhang, S. Zhao, A. Sangiovanni-Vincentelli, K. Keutzer, and B. Gong. (2019). “Domain Randomization and Pyramid Consistency: Simulation-to-Real Generalization without Accessing Target Domain Data”. In: *ICCV*.
- Yun, S., D. Han, S. J. Oh, S. Chun, J. Choe, and Y. Yoo. (2019). “CutMix: Regularization Strategy to Train Strong Classifiers with Localizable Feature”. In: *ICCV*.
- Yurtsever, E., J. Lambert, A. Carballo, and K. Takeda. (2020). “A Survey of Autonomous Driving: Common Practices and Emerging Technologies”. *IEEE Access*. 8: 58443–58469.

- Zang, D., L. Yang, D. Meng, D. Xu, and J. Han. (2017). “SPFTN: A Self-Paced Fine-Tuning Network for Segmenting Objects in Weakly Labelled Videos”. In: *CVPR*.
- Zhan, X., Z. Liu, P. Luo, X. Tang, and C. C. Loy. (2017). “Mix-and-Match Tuning for Self-Supervised Semantic Segmentation”. In: *AAAI*.
- Zhang, J., Z. Ding, W. Li, and P. Ogunbona. (2018a). “Importance Weighted Adversarial Nets for Partial Domain Adaptation”. In: *CVPR*.
- Zhang, L. and X. Gao. (2019). “Transfer Adaptation Learning: A Decade Survey”. arXiv:1903.04687.
- Zhang, M., Y. Zhou, J. Zhao, Y. Man, B. Liu, and R. Yao. (2020a). “A Survey of Semi- and Weakly Supervised Semantic Segmentation of Images”. *Artificial Intelligence Review*. 53: 2402–2417.
- Zhang, Q., J. Zhang, W. Liu, and D. Tao. (2019). “Category Anchor-Guided Unsupervised Domain Adaptation for Semantic Segmentation”. In: *NeurIPS*.
- Zhang, R., P. Isola, and A. A. Efros. (2016a). “Colorful Image Colorization”. In: *ECCV*.
- Zhang, W., Z. Huang, G. Luo, T. Chen, X. Wang, W. Liu, G. Yu, and C. Shen. (2022). “TopFormer: Token Pyramid Transformer for Mobile Semantic Segmentation”. In: *CVPR*.
- Zhang, Y., P. David, H. Foroosh, and B. Gong. (2020b). “A Curriculum Domain Adaptation Approach to the Semantic Segmentation of Urban Scenes”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*. 42(8): 1823–1841.
- Zhang, Y., D. Sidibé, O. Morel, and F. Mériaudeau. (2021). “Deep Multimodal Fusion for Semantic Image Segmentation: A Survey”. *Image and Vision Computing*. 105(104042).
- Zhang, Y., Z. Qiu, T. Yao, C.-W. Ngo, D. Liu, and T. Mei. (2020c). “Transferring and Regularizing Prediction for Semantic Segmentation”. In: *CVPR*.
- Zhang, Y. and A. C. S. Chung. (2018). “Deep Supervision with Additional Labels for Retinal Vessel Segmentation Task”. In: *MICCAI*.

- Zhang, Z., Z. Cui, C. Xu, Z. Jie, X. Li, and J. Yang. (2018b). “Joint Task-Recursive Learning for Semantic Segmentation and Depth Estimation”. In: *ECCV*.
- Zhang, Z., S. Fidler, and R. Urtasun. (2016b). “Instance-Level Segmentation for Autonomous Driving with Deep Densely Connected MRFs”. In: *CVPR*.
- Zhao, H., R. T. des Combes, K. Zhang, and G. Gordon. (2019a). “On Learning Invariant Representation for Domain Adaptation”. In: *ICML*.
- Zhao, H., S. Zhang, G. Wu, J. M. F. Moura, J. P. Costeira, and G. J. Gordon. (2018a). “Adversarial Multiple Source Domain Adaptation”. In: *NeurIPS*.
- Zhao, H., X. Qi, X. Shen, J. Shi, and J. Jia. (2018b). “ICNet for Real-Time Semantic Segmentation on High-Resolution Images”. In: *ECCV*.
- Zhao, H., J. Shi, X. Qi, X. Wang, and J. Jia. (2017). “Pyramid Scene Parsing Network”. In: *CVPR*.
- Zhao, H., Y. Zhang, S. Liu, J. Shi, C. Loy, D. Lin, and J. Jia. (2018c). “PSANet: Point-wise Spatial Attention Network for Scene Parsing”. In: *ECCV*.
- Zhao, S., B. Li, X. Yue, Y. Gu, P. Xu, R. Hu, H. Chai, and K. Keutzer. (2019b). “Multi-source Domain Adaptation for Semantic Segmentation”. In: *NeurIPS*.
- Zhao, S., G. Wang, S. Zhang, Y. Gu, Y. Li, Z. Song, P. Xu, R. Hu, H. Chai, and K. Keutzer. (2020). “Multi-source Distilling Domain Adaptation”. In: *AAAI*.
- Zheng, S., S. Jayasumana, B. Romera-Paredes, V. Vineet, Z. Su, D. Du, C. Huang, and P. H. S. Torr. (2015). “Conditional Random Fields as Recurrent Neural Networks”. In: *ICCV*.
- Zheng, S., J. Lu, H. Zhao, X. Zhu, Z. Luo, Y. Wang, Y. Fu, J. Feng, T. Xiang, P. H. Torr, and L. Zhang. (2021). “Rethinking Semantic Segmentation from a Sequence-to-Sequence Perspective with Transformers”. In: *CVPR*.
- Zheng, S., Y. Song, T. Leung, and I. Goodfellow. (2016). “Improving the Robustness of Deep Neural Networks via Stability Training”. In: *CVPR*.

- Zheng, Z. and Y. Yang. (2020). “Unsupervised Scene Adaptation with Memory Regularization in vivo”. In: *IJCAI*.
- Zheng, Z. and Y. Yang. (2021). “Rectifying Pseudo Label Learning via Uncertainty Estimation for Domain Adaptive Semantic Segmentation”. *International Journal of Computer Vision (IJCV)*. 129: 1106–1120.
- Zhong, E., W. Fan, Q. Yang, O. Verscheure, and J. Ren. (2010). “Cross Validation Framework to Choose Amongst Models and Datasets for Transfer Learning”. In: *PKDD*.
- Zhong, Y., B. Yuan, H. Wu, Z. Yuan, J. Peng, and Y.-X. Wang. (2021). “Pixel Contrastive-Consistent Semi-Supervised Semantic Segmentation”. In: *ICCV*.
- Zhou, B., A. Khosla, A. Lapedriza, A. Oliva, and A. Torralba. (2016). “Learning Deep Features for Discriminative Localization”. In: *CVPR*.
- Zhou, B., H. Zhao, X. Puig, T. Xiao, S. Fidler, A. Barriuso, and A. Torralba. (2018). “A Review of Semantic Segmentation using Deep Neural Networks”. *International Journal of Multimedia Information Retrieval (IJMIR)*. 7: 87–93.
- Zhou, B., H. Zhao, X. Puig, T. Xiao, S. Fidler, A. Barriuso, and A. Torralba. (2019a). “Semantic Understanding of Scenes through the ADE20k Dataset”. *International Journal of Computer Vision (IJCV)*. 127: 302–321.
- Zhou, J., C. Wei, H. Wang, W. Shen, C. Xie, A. Yuille, and T. Kong. (2022). “iBOT:Image BERT Pre-training with Online Tokenizer”. In: *ICLR*.
- Zhou, K., Z. Liu, Y. Qiao, T. Xiang, and C. C. Loy. (2020a). “Domain Generalization: A Survey”. arXiv:2103.02503.
- Zhou, K., Y. Yang, Y. Qiao, and T. Xiang. (2020b). “Domain Adaptive Ensemble Learning”. arXiv:2003.07325.
- Zhou, T., S. Ruan, and S. Canu. (2019b). “A Review: Deep Learning for Medical Image Segmentation using Multi-modality Fusion”. *Array*. 3-4(100004).
- Zhou, Y., H. Xu, W. Zhang, B. Gao, and P.-A. Heng. (2021). “ C^3 -SemiSeg: Contrastive Semi-supervised Segmentation via Cross-set Learning and Dynamic Class-balancing”. In: *ICCV*.

- Zhou, Z.-H. and M. Li. (2005). “Tri-training: Exploiting Unlabeled Data using Three Classifiers”. *IEEE Transactions on Knowledge and Data Engineering*. 11(17): 1529–1541.
- Zhu, J.-Y., T. Park, P. Isola, and A. A. Efros. (2017). “Unpaired Image-to-Image Translation Using Cycle-Consistent Adversarial Networks”. In: *ICCV*.
- Zhu, X., H. Zhou, C. Yang, J. Shi, and D. Lin. (2018). “Penalizing Top Performers: Conservative Loss for Semantic Segmentation Adaptation”. In: *ECCV*.
- Zhu, Y., Z. Zhang, C. Wu, Z. Zhang, T. He, H. Zhang, R. Manmatha, M. Li, and A. Smola. (2021). “Improving Semantic Segmentation via Efficient Self-Training”. *IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)*.
- Zhu, Y., F. Zhuang, and D. Wang. (2019). “Aligning Domain-Specific Distribution and Classifier for Cross-Domain Classification from Multiple Sources”. In: *AAAI*.
- Zou, Y., Z. Yu, B. V. Kumar, and J. Wang. (2018). “Unsupervised Domain Adaptation for Semantic Segmentation via Class-Balanced Self-Training”. In: *ECCV*.
- Zou, Y., Z. Yu, X. Liu, B. V. K. V. Kumar, and J. Wang. (2019). “Confidence Regularized Self-Training”. In: *ICCV*.
- Zurbrügg, R., H. Blum, C. Cadena, R. Siegwart, and L. Schmid. (2022). “Embodied Active Domain Adaptation for Semantic Segmentation via Informative Path Planning”. arXiv:2203.00549.