

A Downstream Task Informed Domain Adaptation GAN for Semantic Segmentation

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Abstract and Contribution

We focus on understanding complex visual street scenes with the help of pixelwise classification (semantic segmentation). As labeled data is often rare, we consider computer simulations of urban scenes, where we can generate arbitrary many labeled samples with no or little cost and train a network on the fully controllable domain.

Neural networks may perform poorly when used in domains with a different underlying data distribution. This phenomenon is called domain gap. To solve semantic segmentation in other domains like the real world where labels are hard to obtain, we aim to mitigate the gap via style transfer and guidance of the generator towards the downstream task.

Our main contributions in this work are:

- Feasibility analysis of exploiting a fully controllable domain
- A semi-supervised method to guide the generator to a downstream task without retraining the downstream task network
- A more unbiased domain gap analysis by using a from scratch trained semantic segmentation network.

Concept of Generative Adversarial Networks

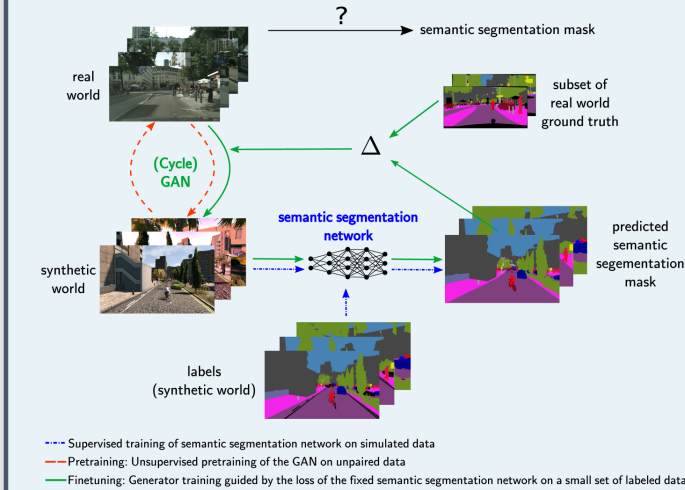


References

- [1] Jülich Supercomputing Centre, "JUWELS: Modular Tier-0/1 Supercomputer at the Jülich Supercomputing Centre," *Journal of large-scale research facilities*, vol. 5, no. A135, 2019.
- [2] X. Mao, Q. Li, H. Xie, R. Y. K. Lau, Z. Wang, and S. P. Smolley, "Least squares generative adversarial networks," *ICCV*, pp. 2813–2821, 2017.
- [3] L.-C. Chen, G. Papandreou, F. Schroff, and H. Adam, "Rethinking Atrous Convolution for Semantic Image Segmentation," *arXiv:1706.05587 [cs]*, Dec. 2017.
- [4] G. Ros, L. Sellart, J. Materzynska, D. Vazquez, and A. Lopez, "The synthia dataset: A large collection of synthetic images for semantic segmentation of urban scenes," in *CVPR*, 2016.
- [5] M. Cordts, M. Omran, S. Ramos, T. Rehfeld, M. Enzweiler, R. Benenson, U. Franke, S. Roth, and B. Schiele, "The cityscapes dataset for semantic urban scene understanding," in *CVPR*, 2016.
- [6] Y.-H. Tsai, W.-C. Hung, S. Schuster, K. Sohn, M.-H. Yang, and M. Chandraker, "Learning to adapt structured output space for semantic segmentation," in *CVPR*, 2018.

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Concept of Task Informed GAN (ours)



Extended Generator Loss \mathcal{L}_{G_S}

$$\mathcal{L}_{G_S}(I) = (1 - \alpha) \underbrace{\left\| D_S(G_S(I)) - 1 \right\|^2}_{\text{original LSGAN generator loss [2]}} + \alpha \underbrace{\left(\gamma CE(f(G_S(I)), l_r) \right)}_{\text{Cross entropy loss}},$$

with generator G_S from the real domain \mathcal{R} to the synthetic domain \mathcal{S} , discriminator D_S , semantic segmentation network f , real world image I with label l_r and $G_S(I)$ its GAN transformed version.

Domain Gap

Our setup: DeepLabV3 [3] with ResNet101 as backbone trained from scratch on only synthetic data (Synthia [4]).

input	pretrained	mIoU
Synthia [4] valset	no	0.648
Cityscapes [5] valset	no	0.099
Cityscapes [5] valset	yes	0.386 [6]

→ indication that backbones pre-trained on real world classification tasks are biased towards the real domain.

GAN and Segmentation Results (based on Synthia)

