

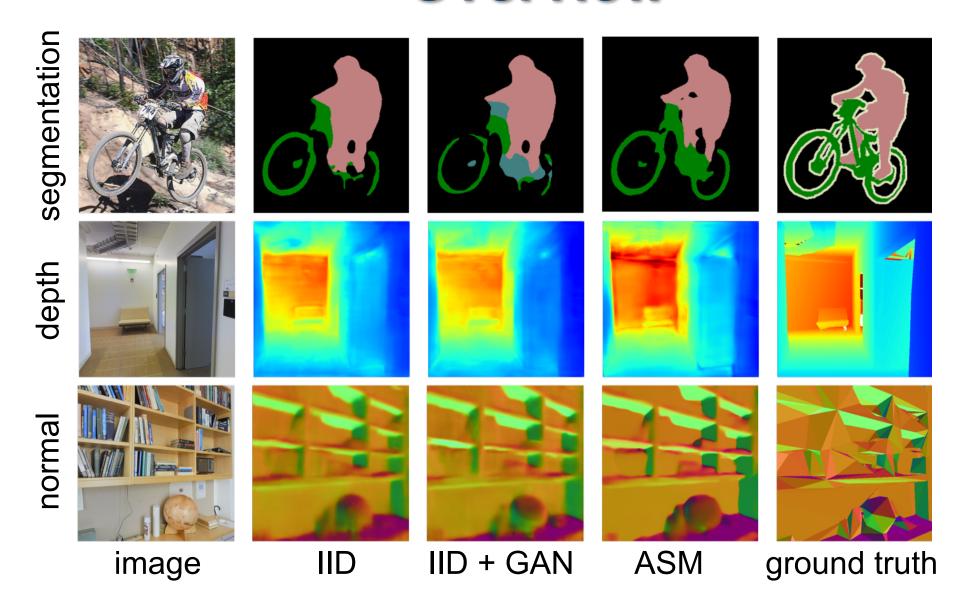


Adversarial Structure Matching for Structured Prediction Tasks Jyh-Jing Hwang*, Tsung-Wei Ke*, Jianbo Shi, Stella X. Yu

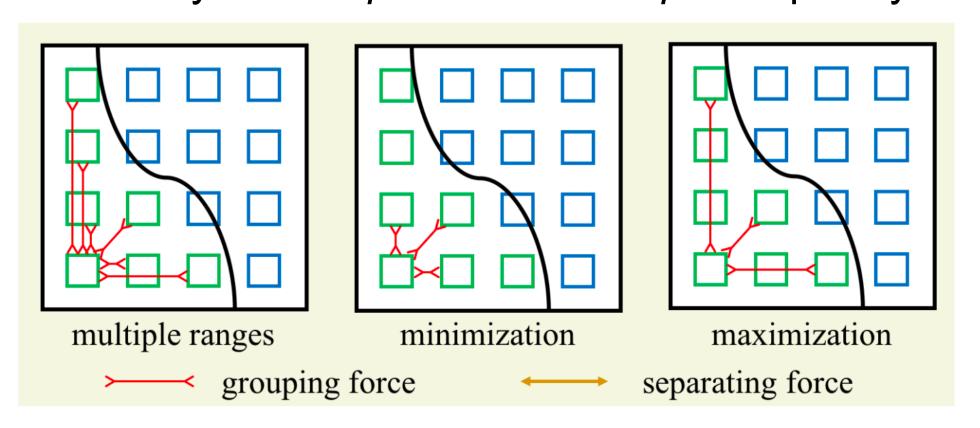
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Overview

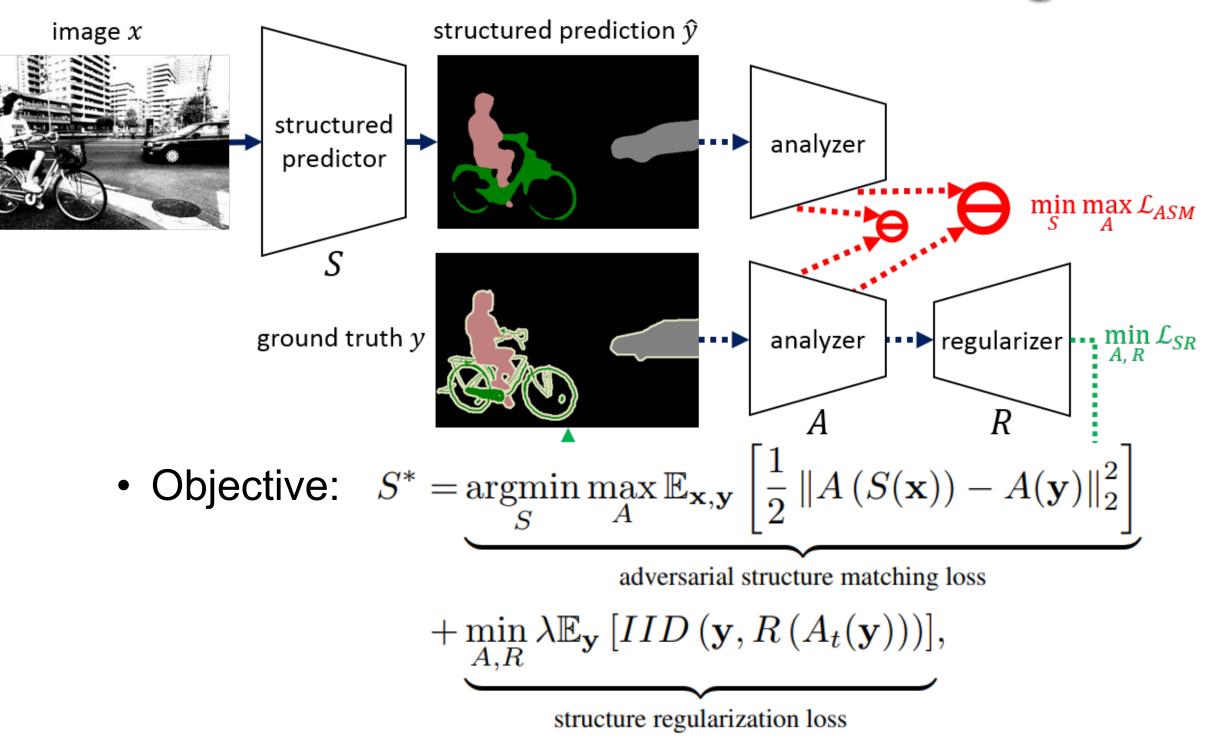


- Structured predictions output correlated 2D masks, including semantic segmentation, depth estimation, surface normal prediction, etc..
- IID (softmax / L2) is the most common approach, which ignores label correlations among pixels.
- GAN is a prior based structural model, which encodes relationships in a one-to-set mapping.
- Our ASM adversarially matches *multiscale structures* in the label space, featuring:
- 1. Adaptive structure prior
- 2. Instance specificity
- 3. Generalizability.
- Adaptive Affinity Fields (2nd order regularization): AAF only selects *pixel relationships* adaptively.

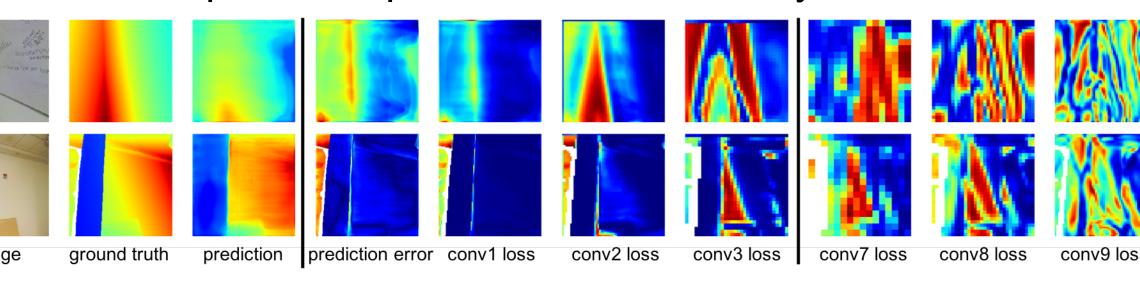


Adaptive Affinity Fields for Semantic Segmentation, ECCV 2018.

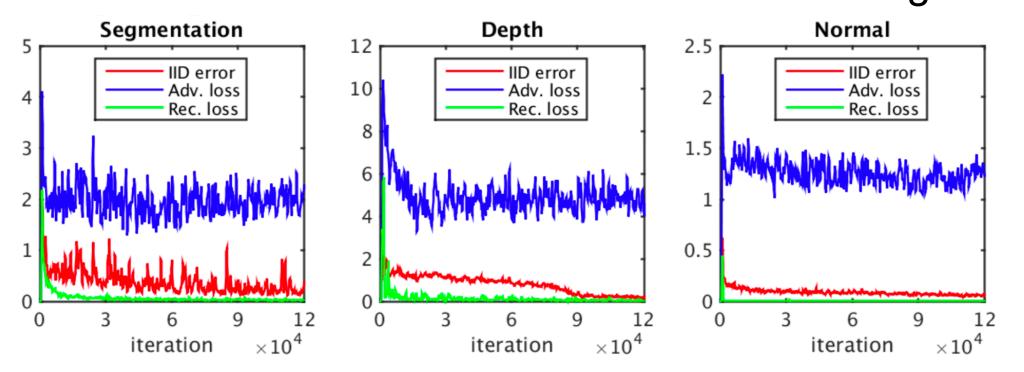
Adversarial Structure Matching



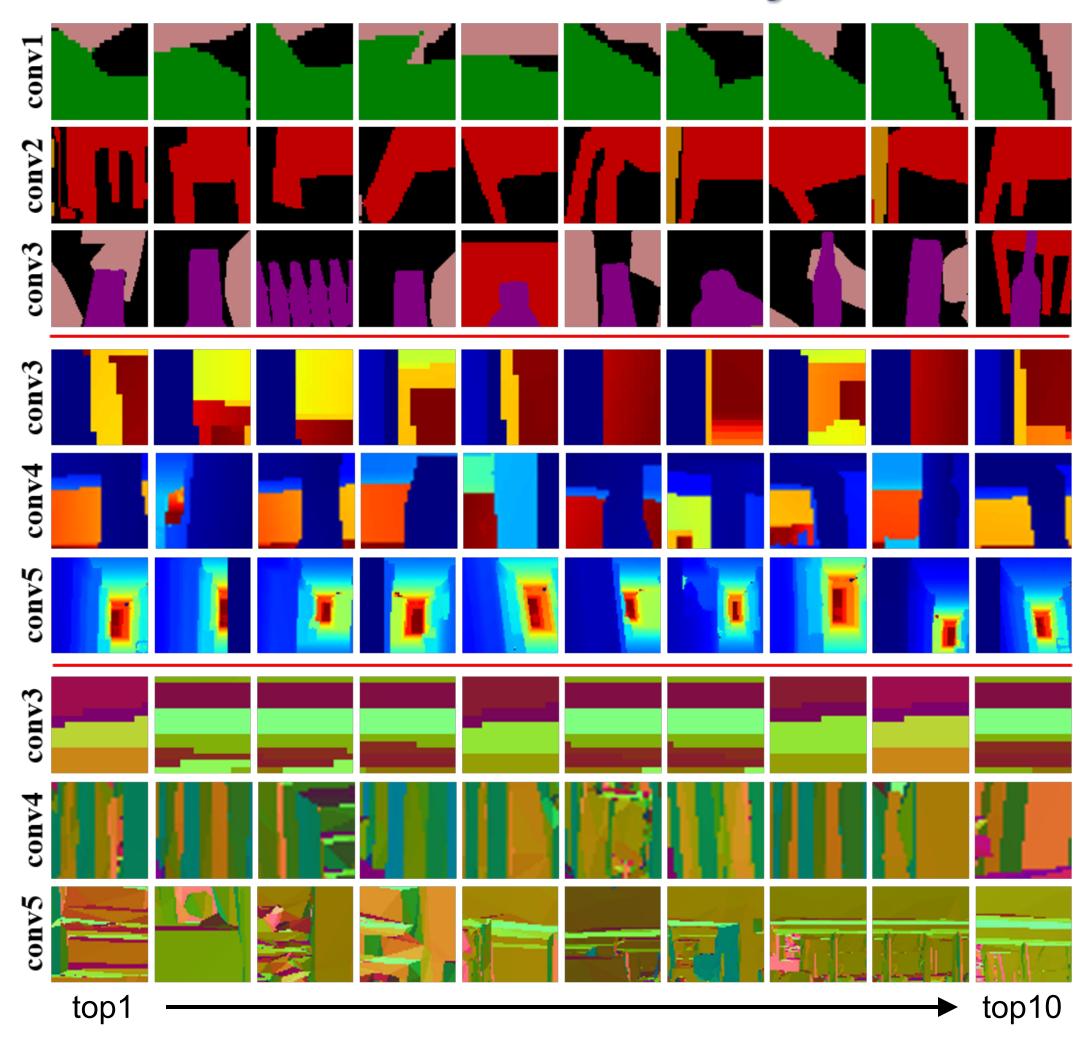
Sampled loss patterns in different layers:



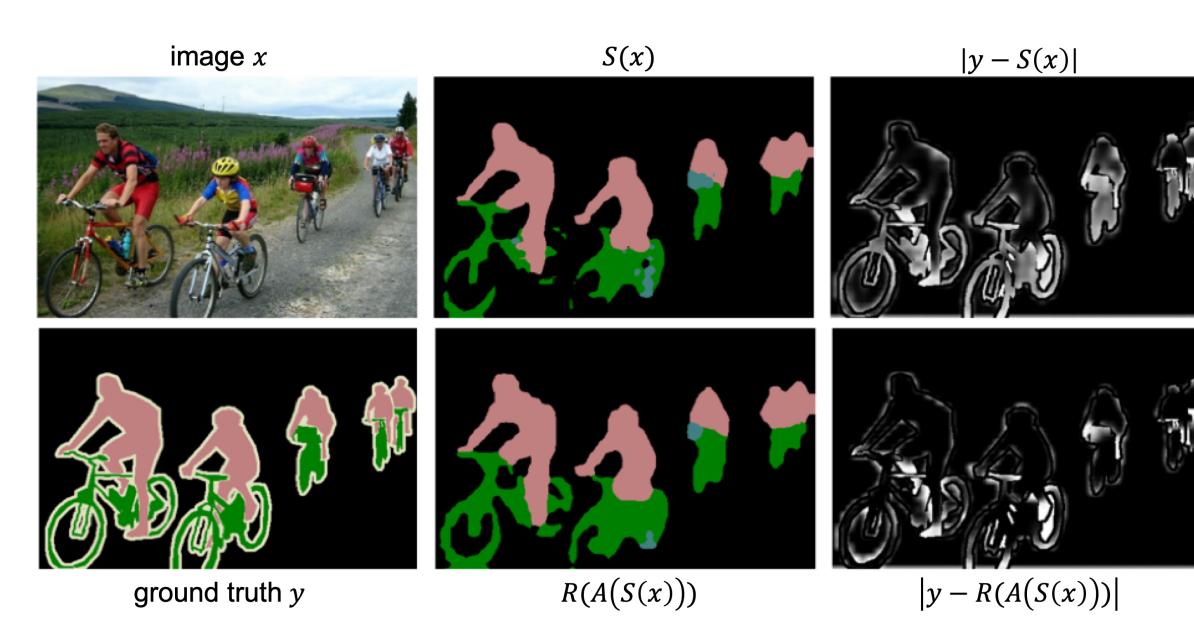
- Analyzer A is trained to maximize the multiscale structural mistakes of S. → Hard negative mining.
- *Predictor S* is trained to minimize the same error.
- Regularizer R ensures that A also forms a good basis for reconstructing the ground truth.
- Assuming infinite capacity for S and A, we proved $S^*(x) = y$ and $V(S^*, A^*) = 0$. (Nash equilibrium)
- ASM retains critical assessments thru training:



What's Learned in Analyzer A?



- A encodes multiscale pixel relationships, e.g., person riding bike, hand picking up bottle, etc..
- A and R learn to complete shapes while
 A adaptively refines the focus of supervision.



Experimental Results

