Proposal-based Instance Segmentation with Point Supervision











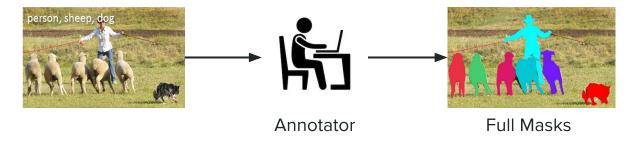
Laradji et al. accepted at ICIP2020

Instance Segmentation with Point Supervision Motivation



Instance Segmentation with Point Supervision Motivation

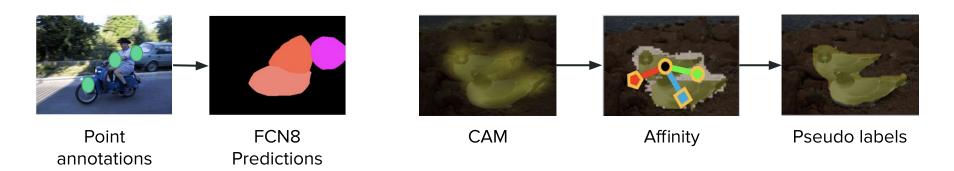
Per-pixel labels require 1.5 hours/image



Point-level labels require few seconds/image



Related work

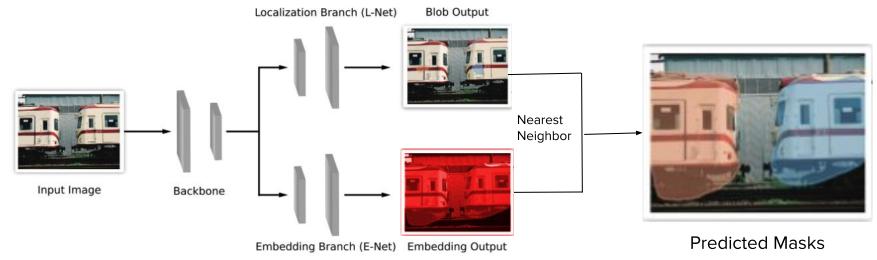


What's the point [ECCV 2016]

Learning semantic affinity [CVPR 2018]

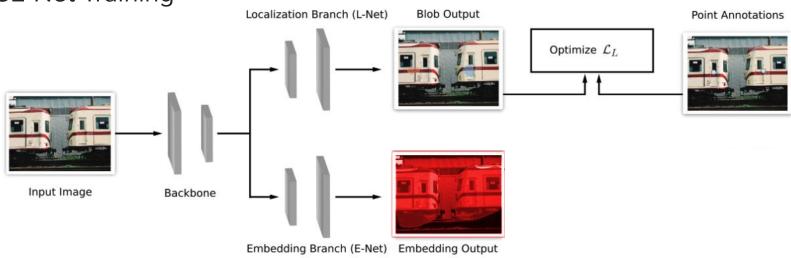
However, these are weakly supervised semantic segmentation methods

WISE-Net Prediction



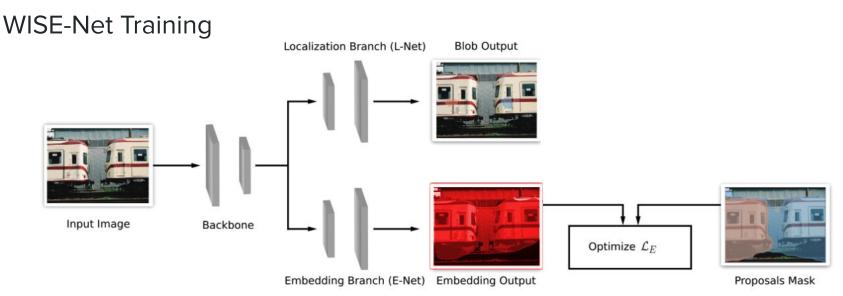
- Locate the objects with the localization branch
- Obtain the masks with the embedding branch

WISE-Net Training



LCFCN loss to predict the object locations:

$$\mathcal{L}_{L} = \underbrace{\mathcal{L}_{I}(S,T)}_{\text{Image-level loss}} + \underbrace{\mathcal{L}_{P}(S,T)}_{\text{Point-level loss}} + \underbrace{\mathcal{L}_{S}(S,T)}_{\text{Split-level loss}} + \underbrace{\mathcal{L}_{F}(S,T)}_{\text{False positive loss}}$$

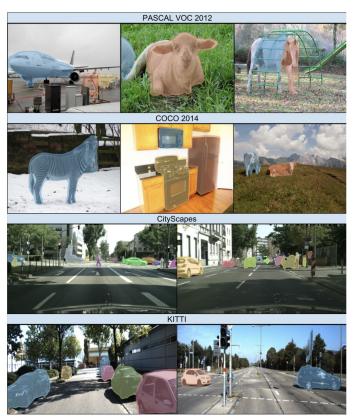


Similarity loss makes pixels of the same object have similar embeddings:

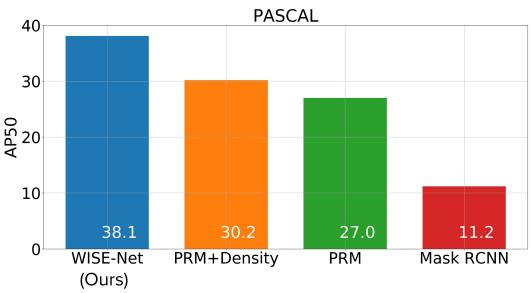
$$\mathcal{L}_{E} = -\sum_{(i,j)\in P} \left[\mathbb{1}_{\{y_{i}=y_{j}\}} \log S(E_{i}, E_{j}) + \mathbb{1}_{\{y_{i}\neq y_{j}\}} \log \left(1 - S(E_{i}, E_{j})\right) \right]$$

where few points are sampled from the proposals and from the background regions

Results



Same labeling effort are given for each of these methods



Higher AP50 is better

Conclusions

- A novel framework instance segmentation with point supervision
- Outperforms fully- and weakly-supervised methods for fixed annotation budget
- Established a first strong baseline for the problem setup
- Check out our code in https://github.com/lssamLaradji/wisenet

Instance Segmentation with Point Supervision Real life applications



Instance Segmentation with Point Supervision Real life applications







