



Motivation Success Factors of Deep Learning



- availability of hardware for massive parallel computations
- large-scale labeled datasets
 - ImageNet dataset contains more than 14 mio labeled images
 - ☐ Youtube-8M dataset contains more than 7 mio labeled videos



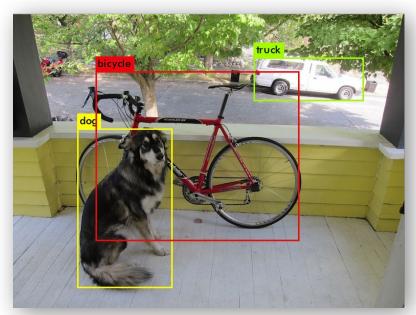
Motivation Object Detection



- recent breakthroughs [Redmon16, Liu16, Ren15] use fully annotated datasets
 - labels for locations of objects and classes of objects



```
[
    "class": "dog",
    "box": [200, 50, 400, 100]
},
{
    "class": "bicycle",
    "box": [100, 60, 350, 225]
},
{
    "class": "truck",
    "box": [75, 200, 150, 300]
}
]
```

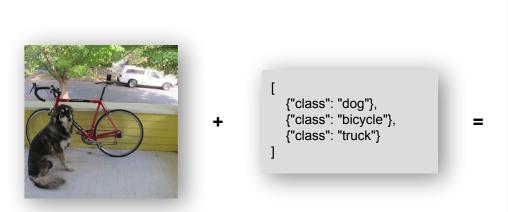


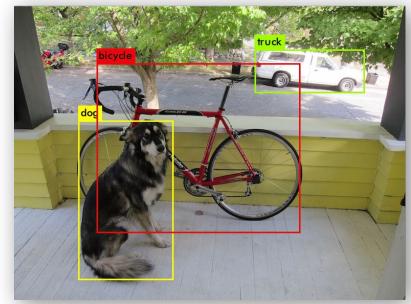
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Motivation Weakly Supervised Object Detection



- recent methods [Wei18, Tang18] only use class labels
 - leverage implicit localization capability of feature extractor





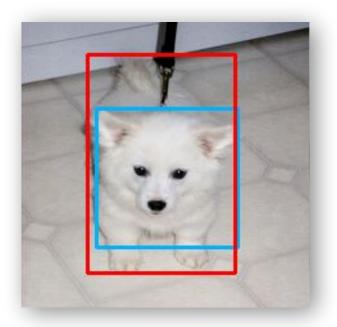
Bartz, Yang, Bethge, Meinel

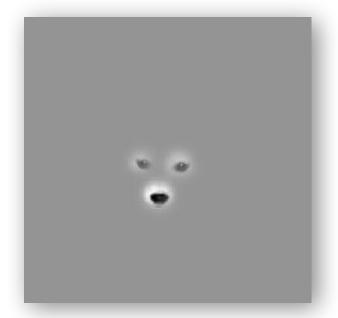
Chart 4

Motivation Weakly Supervised Object Detection



current weakly supervised methods have inherent problems



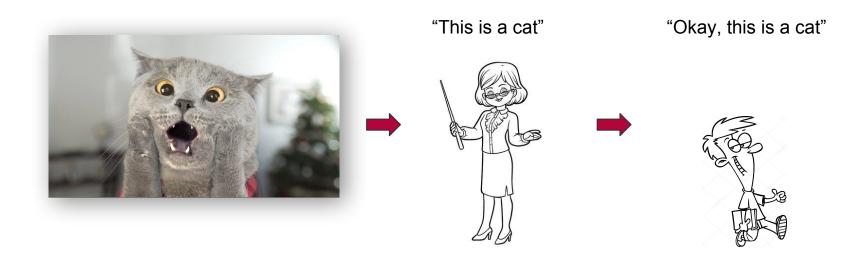


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Motivation Teacher Student Networks



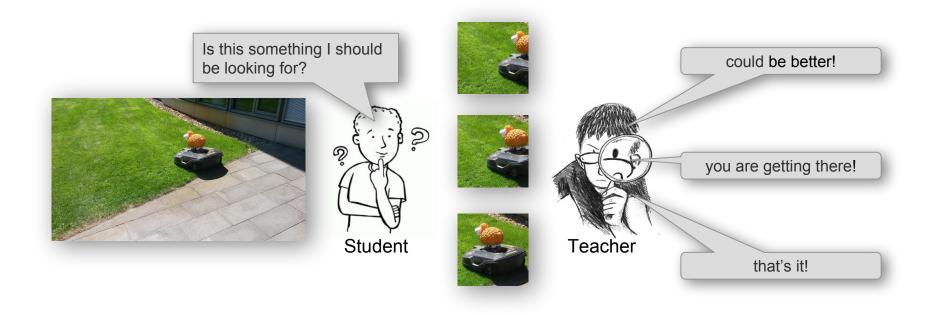
knowledge transfer [Hinton15, Chen16] between networks does not need labels for training the student



LoANs: Localizer Assessor Networks

Motivation Idea

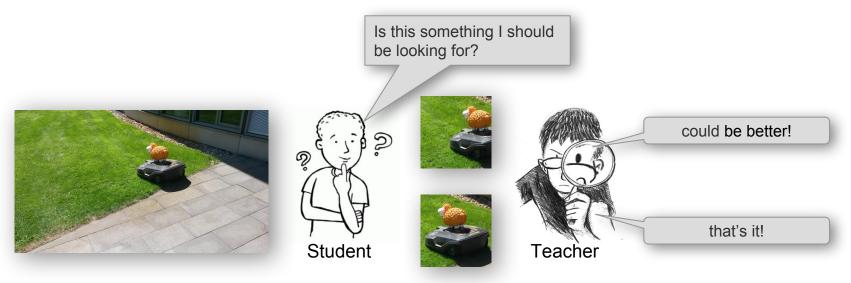




Motivation Idea



- no labels for bounding boxes necessary
- no model pre-trained on ImageNet necessary
- can use artificial data for training teacher model

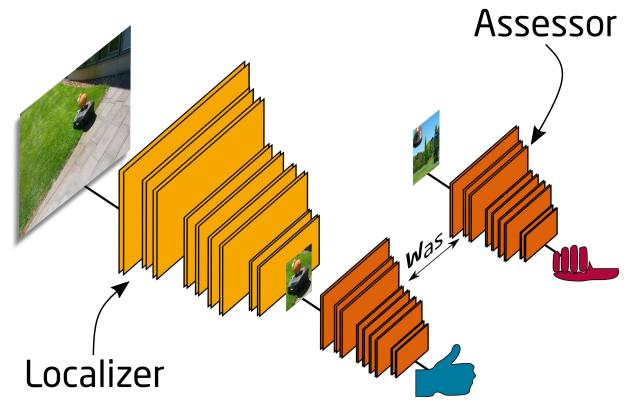


LoANs: Localizer Assessor Networks

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Proposed System Localizer Assessor Networks





Proposed System Assessor



- predict intersection over union (IOU) of image crop and shown object
- trained on synthetically generated data
- needs background and template images





Step 0: select a background image



Step 1: place template image at random location in background image



Step 2: find a box with desired intersection over union



Step 3: crop box and resize image

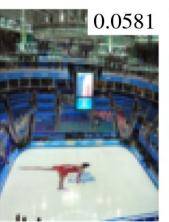
Proposed System Assessor

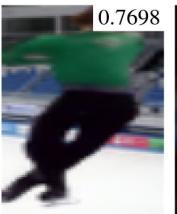


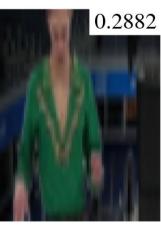
 save image and intersection over union of crop with bounding box of object











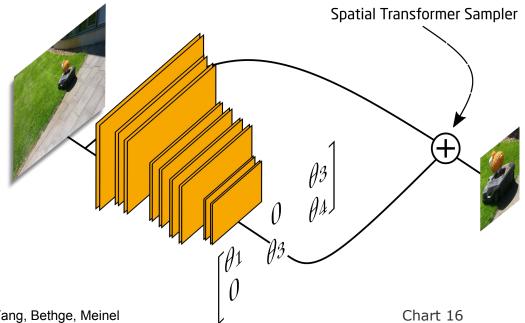
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Proposed System Localizer



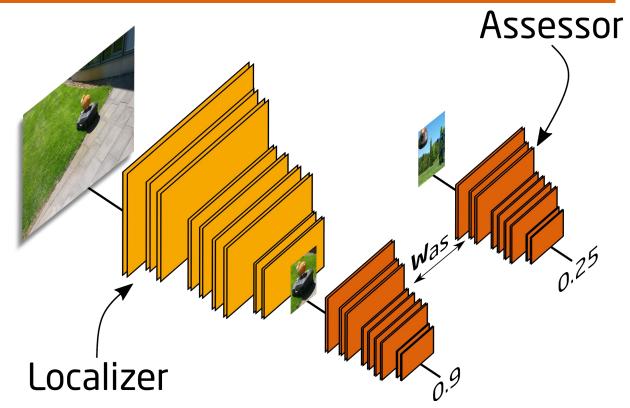
- predict a image region that is likely to contain target object
- crop image region with a spatial transformer [Jaderberg15]
- trained on unlabeled data
- entirely supervised by assessor



LoANs: Localizer Assessor Networks

Proposed System Training of Both Networks

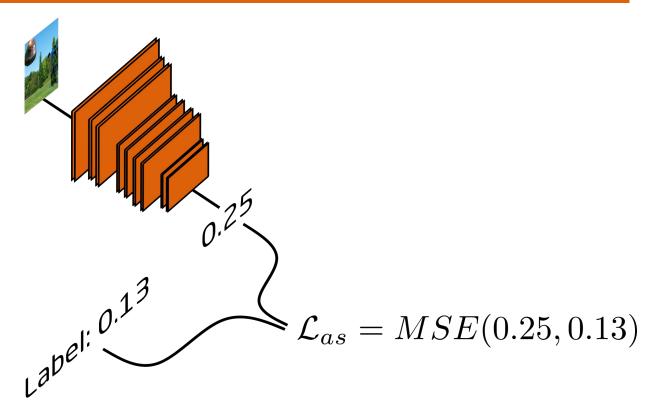




LoANs: Localizer Assessor Networks

Proposed System Training of Assessor

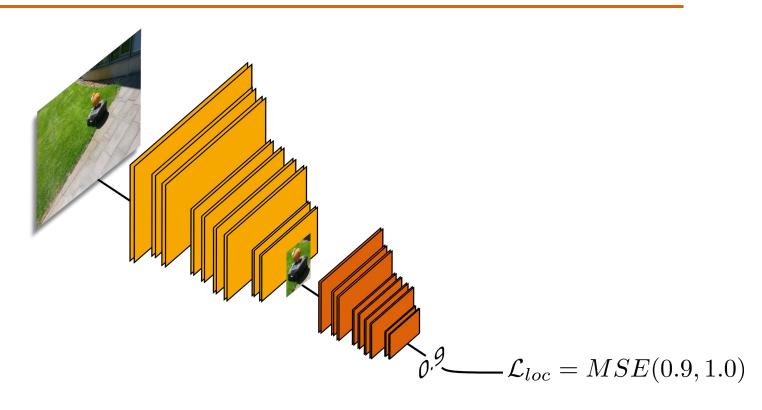




LoANs: Localizer Assessor Networks

Proposed System Training of Localizer





Experiments Sheep Dataset



- 8,320 fully annotated images for training localizer
 - full annotation enables comparison to fully supervised approach
- 10,000 images for training assessor

Method	224 x 224	300 x 300	512 x 512
SSD [Liu16]	-	0.887	0.969
ResNet-18	0.887	0.937	0.967
ResNet-50	0.959	0.958	0.976

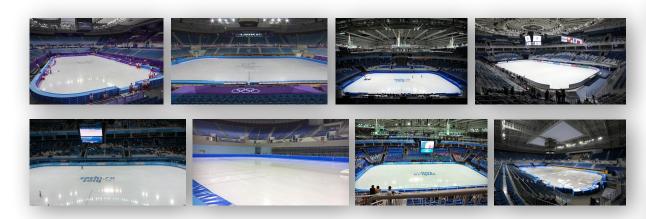


Experiments Figure Skating Dataset



 8 background images and 25 template images are enough for generation of assessor dataset

4 YouTube videos are enough to create train dataset for localizer





LoANs: Localizer Assessor Networks

Experiments Figure Skating Dataset



- YouTube videos contain a lot of noisy images
- experimented with noisy data and without noise
 - dataset without noise only contains 48% of the number of original images



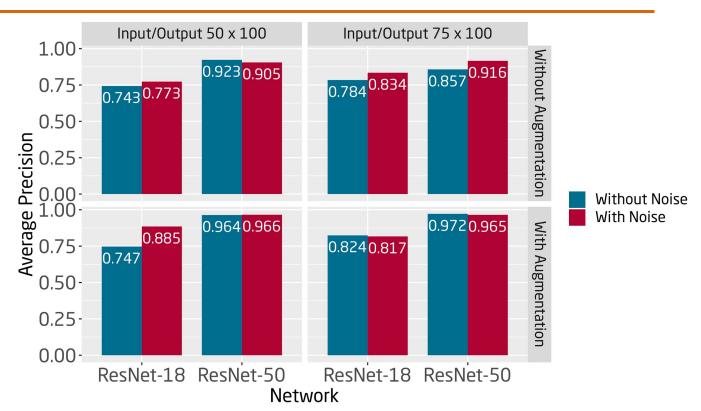
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Chart 23

Experiments Results on Figure Skating Dataset



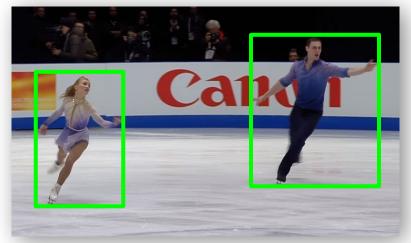


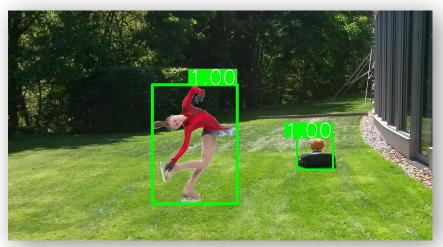
Limitations and Future Work



- approach only works for one class at a time
- only works with images containing a single object
- we have no means to determine whether system detected object or not

What we would like to have in the future:





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Chart 25

Conclusion



presented a novel approach for weakly supervised detection

should be simple and cost efficient to create training data for

specialized systems

code, models and datasets are available online:

https://github.com/Bartzi/loans



Thank you for your attention

References



- [Redmon16] Redmon, Joseph, et al. "You only look once: Unified, real-time object detection." Proceedings of the IEEE conference on computer vision and pattern recognition. 2016.
- [Liu16] Liu, Wei, et al. "Ssd: Single shot multibox detector." European conference on computer vision. Springer, Cham, 2016.
- [Ren15] Faster R-CNN: Towards Real-Time Object Detection with Region Proposal Networks
- [Wei18] Wei, Yunchao, et al. "TS2C: tight box mining with surrounding segmentation context for weakly supervised object detection." European Conference on Computer Vision. Springer, Cham, 2018.

References



- [Tang18] Tang, Peng, et al. "Weakly supervised region proposal network and object detection." Proceedings of the European Conference on Computer Vision (ECCV). 2018.
- [Hinton15] Hinton, Geoffrey, Oriol Vinyals, and Jeff Dean. "Distilling the Knowledge in a Neural Network." NIPS Deep Learning and Representation Learning Workshop. 2015.
- [Chen16] Chen, Tianqi, Ian Goodfellow, and Jonathon Shlens.
 "Net2net: Accelerating learning via knowledge transfer."
 International Conference on Learning Representations. 2016.

References



[Jaderberg15] - Jaderberg, Max, Karen Simonyan, and Andrew Zisserman. "Spatial transformer networks." Advances in neural information processing systems. 2015.