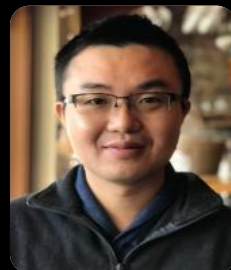
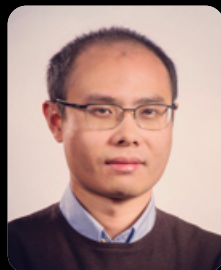
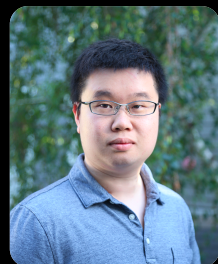


# UFO<sup>2</sup>: A Unified Framework towards Omni-supervised Object Detection

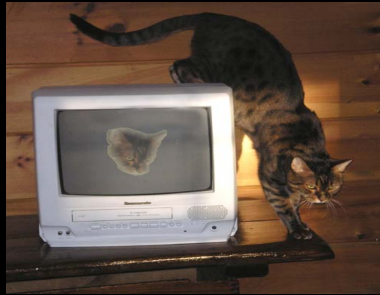
Zhongzheng Ren, Zhiding Yu, Xiaodong Yang, Ming-Yu Liu,  
Alexander G. Schwing, Jan Kautz

ECCV 2020

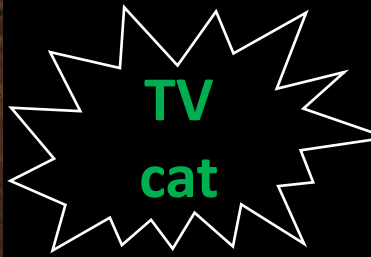


# Omni-supervised Object Detection

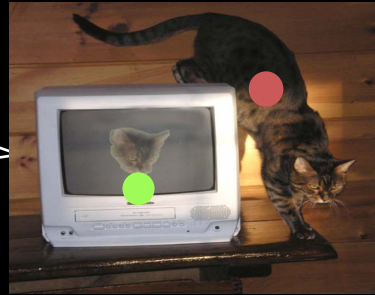
unlabeled



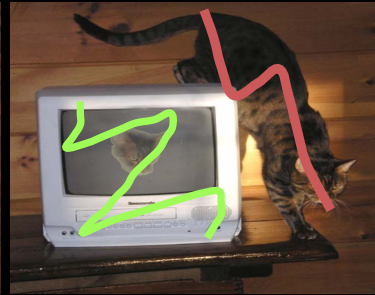
tags



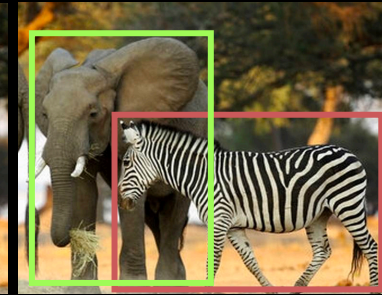
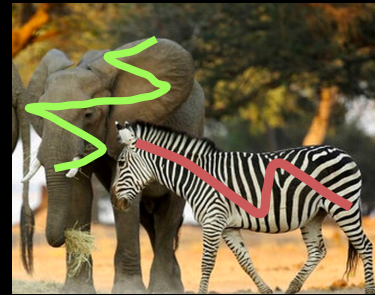
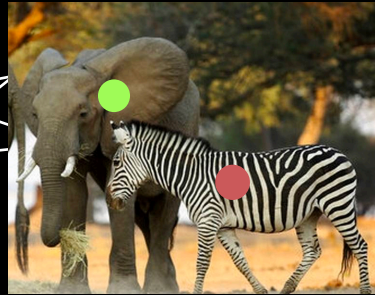
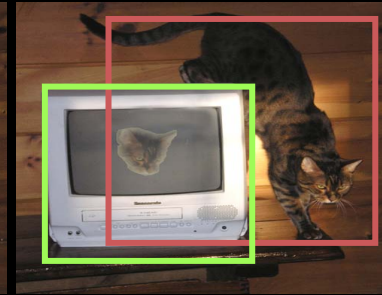
points



scribbles



boxes



~~weakly-supervised~~

commonly used in

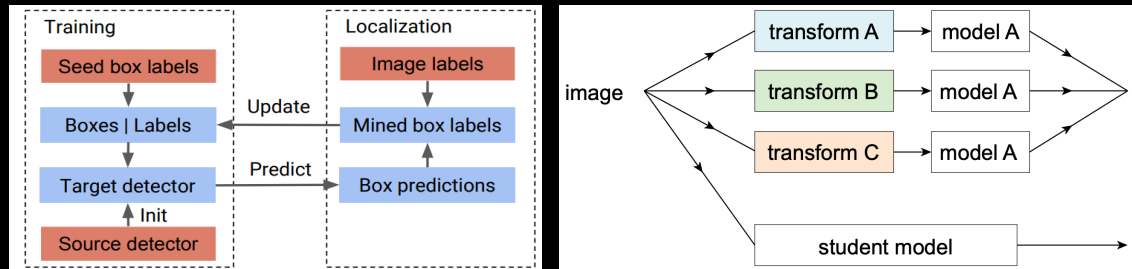
~~strongly-supervised~~

e.g., segmentation

omni-supervised

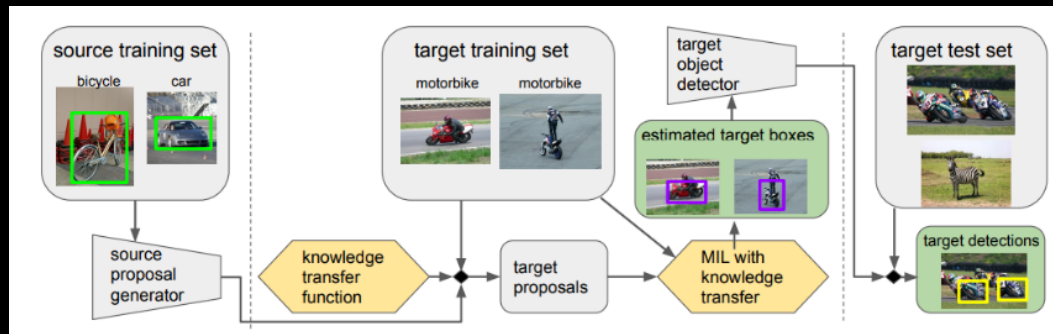


# Omni-supervised Object Detection



(Gao et al., 2018)

(Radosavovic et al., 2019)



(Uijlings et al., 2018)

## Semi-supervised Object Detection

### Prior work

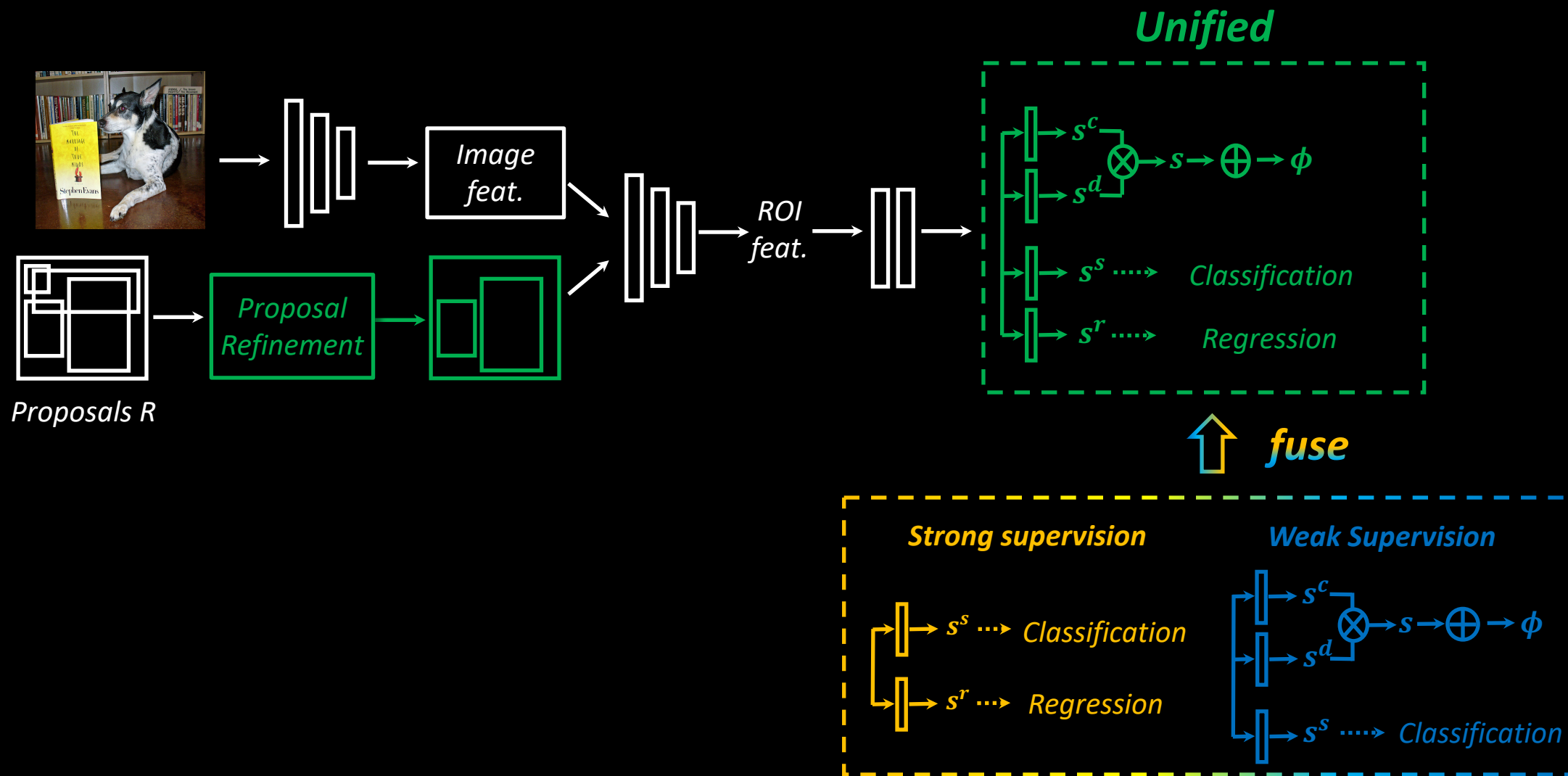
- Stage-wise training; pipelined
- Require some strong labels

### Ours

- Unified
- Strong labels are not necessary
- More labels supported



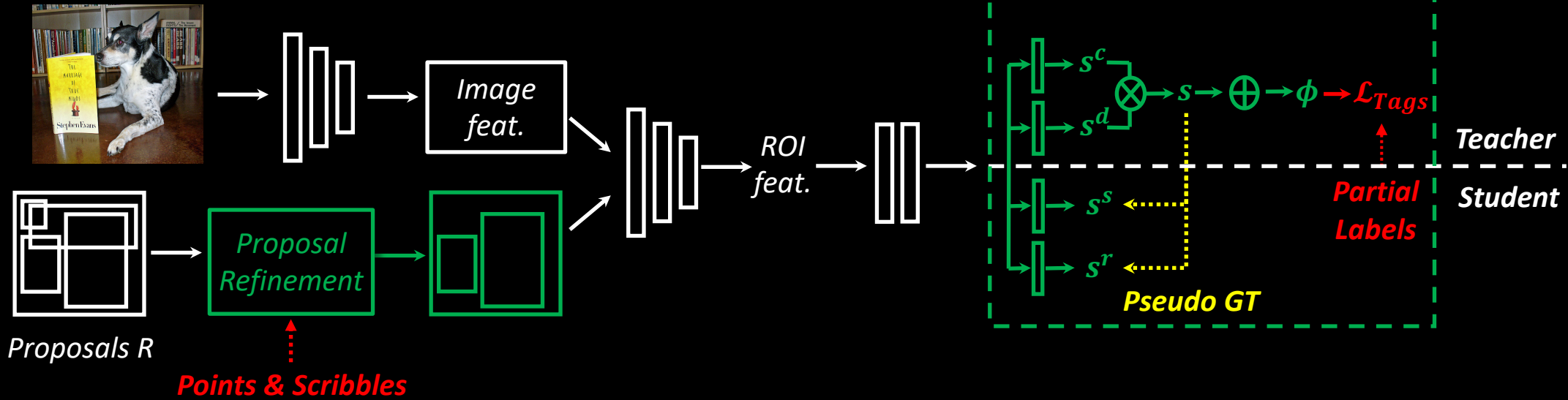
# UFO<sup>2</sup>: a Unified Framework





# Partial Labels (tags, points, scribbles)

Unified



No GT boxes available

Teacher heads:

- image-level multi-label classification ( $\mathcal{L}_{Tags}$ )

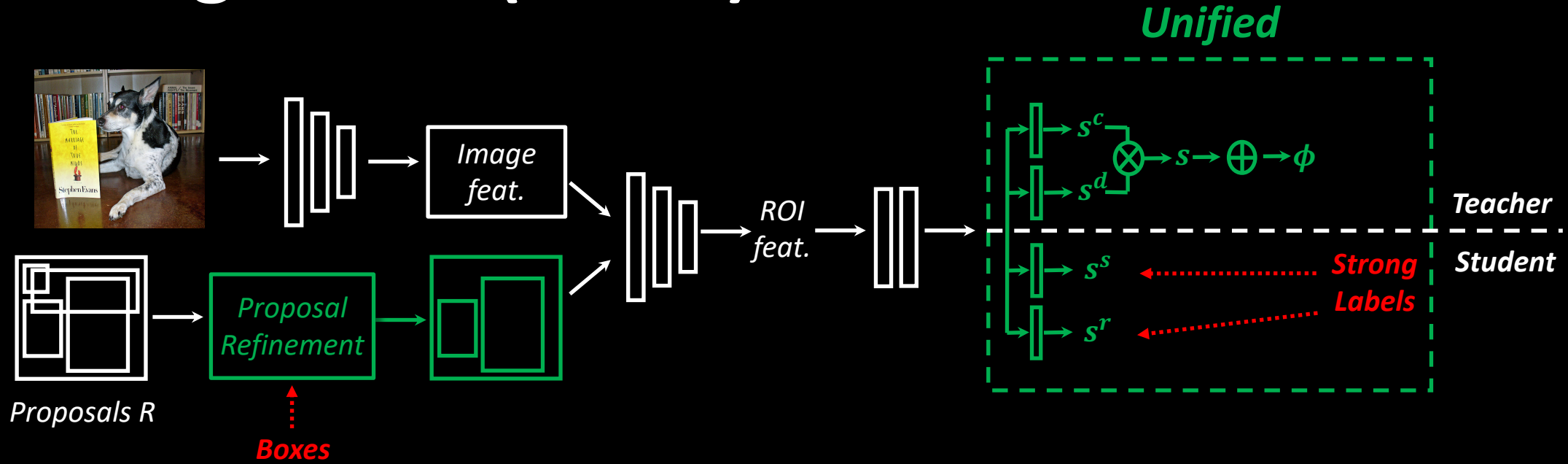
Generate pseudo-GT online

Student heads:

- RoI classification
- RoI regression



# Strong Labels (boxes)



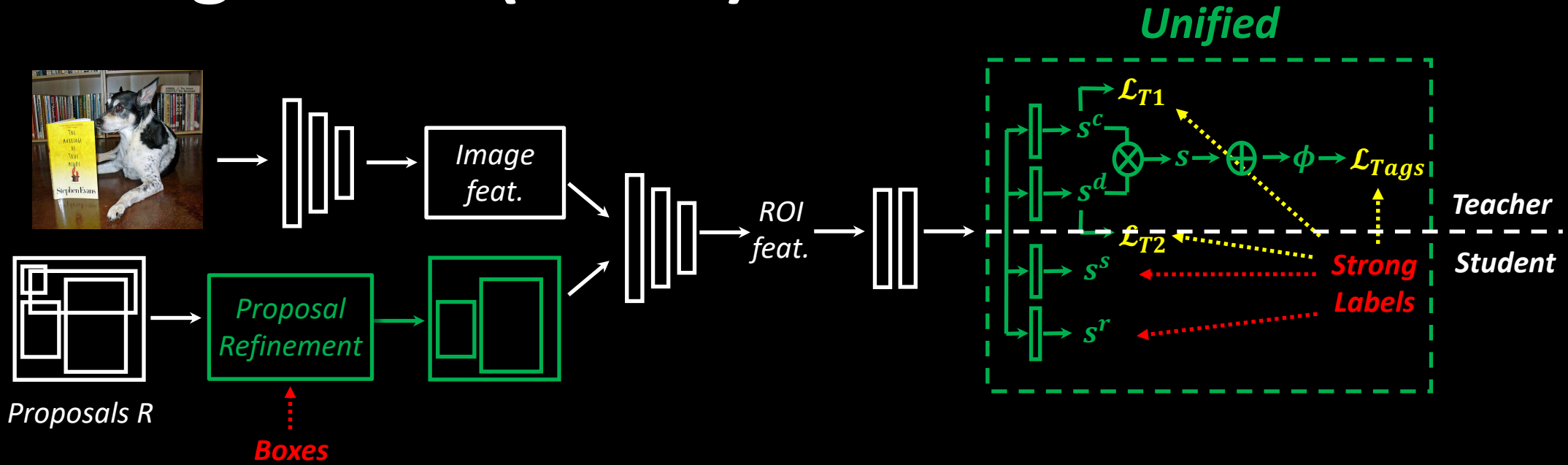
Naïve solution: directly supervise student heads using GT boxes



**Issue: Weak Teacher & Strong Students**



# Strong Labels (boxes)

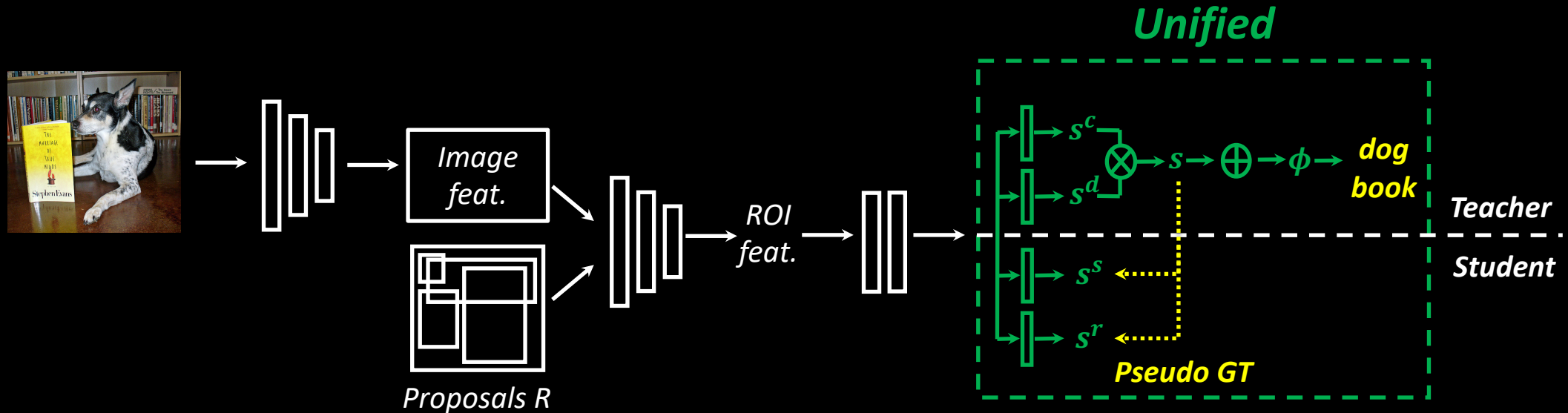


## Make Teacher Great Again!

- Image-level multi-label classification ( $\mathcal{L}_{Tags}$ )
- RoI classification ( $\mathcal{L}_{T1}$ )
- RoI objectness regularization ( $\mathcal{L}_{T2}$ )



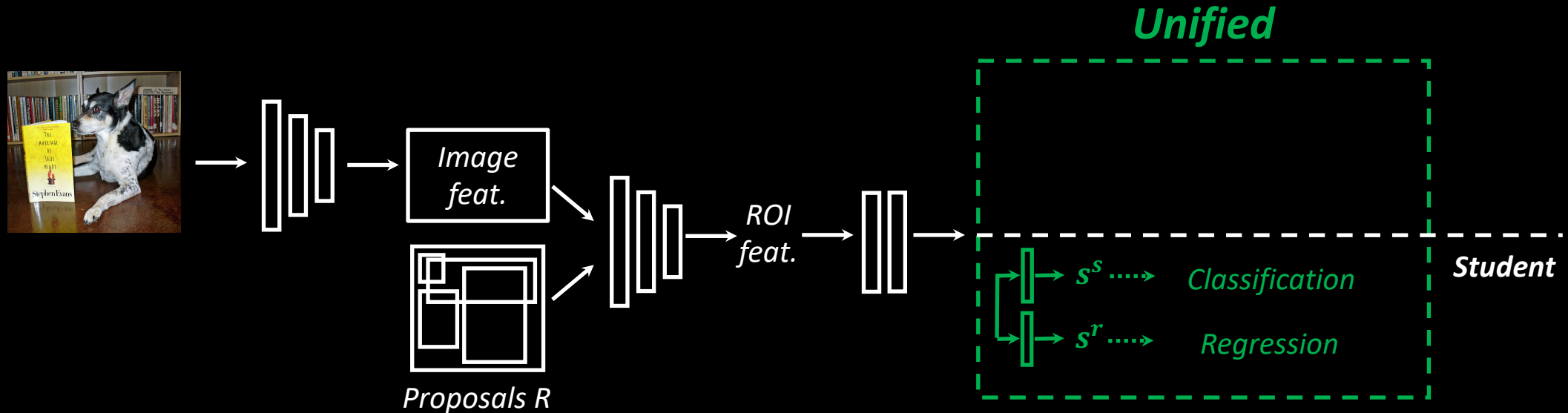
# Unlabeled Data



- Take confident classes from teacher's prediction
- Then follow "Tags" setting



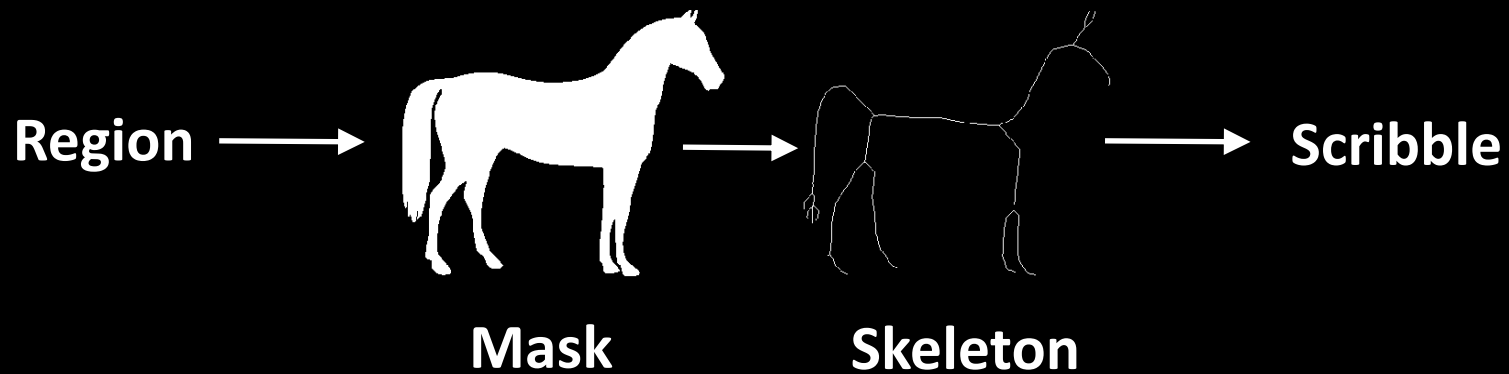
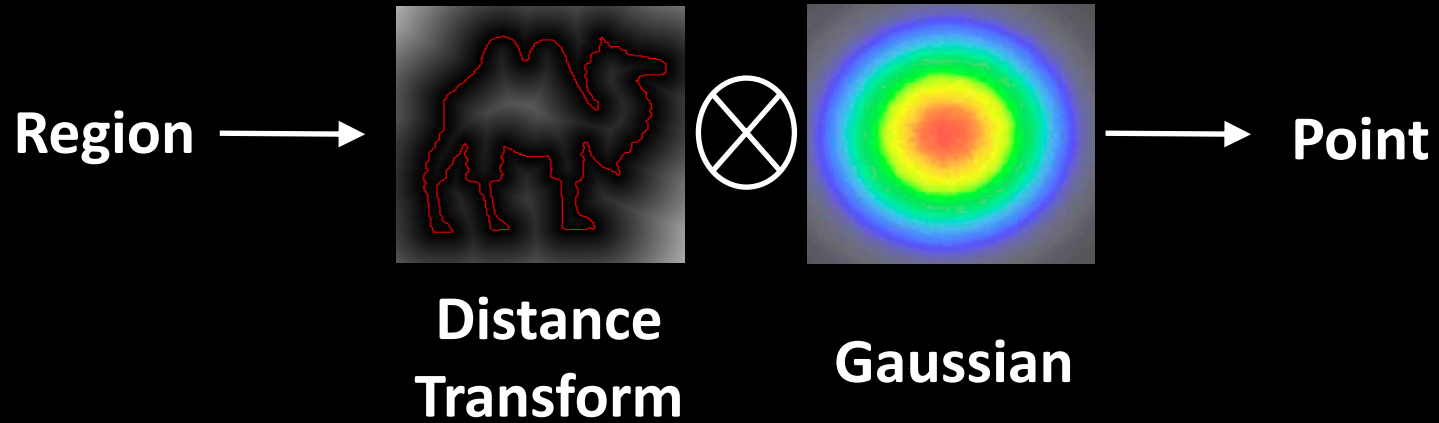
# Inference



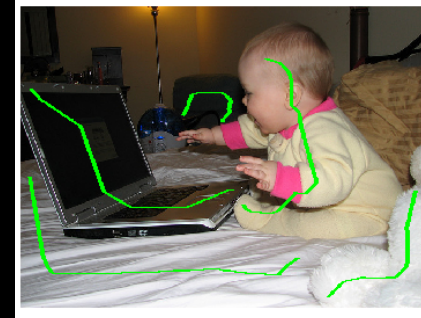
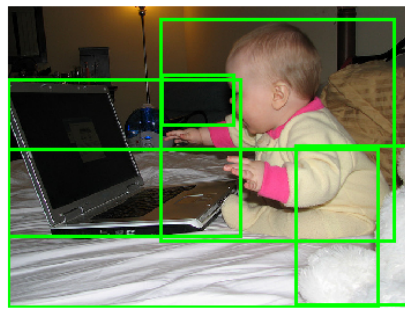
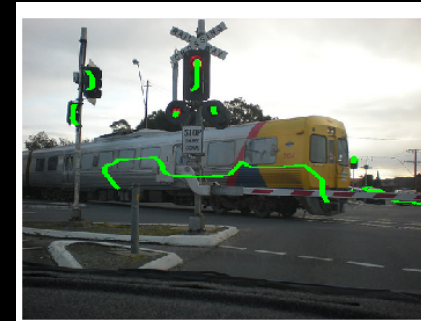
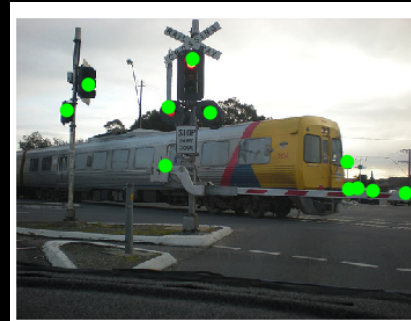
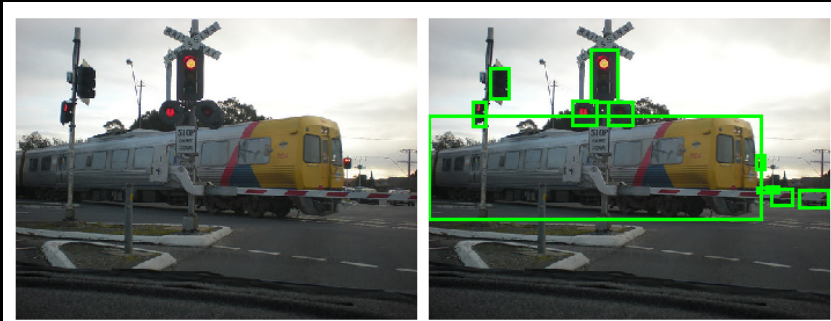
- Use only the student heads
- As efficient as standard supervised detectors



# Dataset: Partial Labels Simulation



# Dataset: Partial Labels Simulation



COCO images & boxes

points

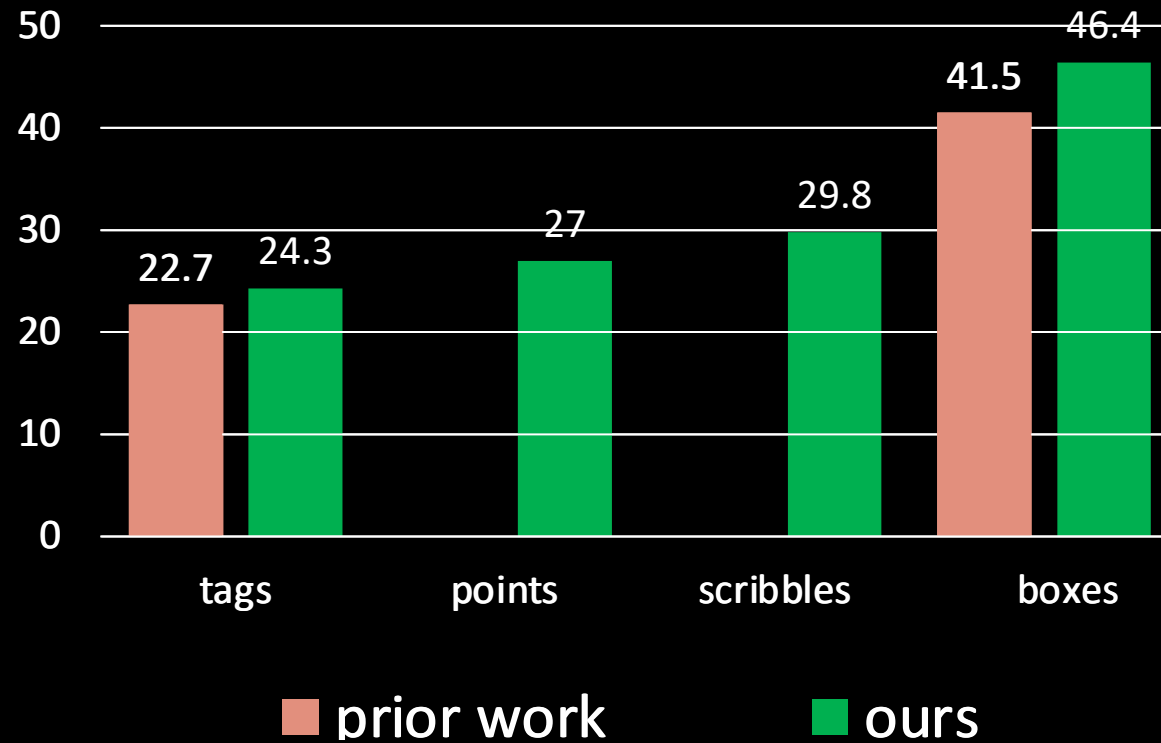
scribbles





# Experiments: Train from Scratch

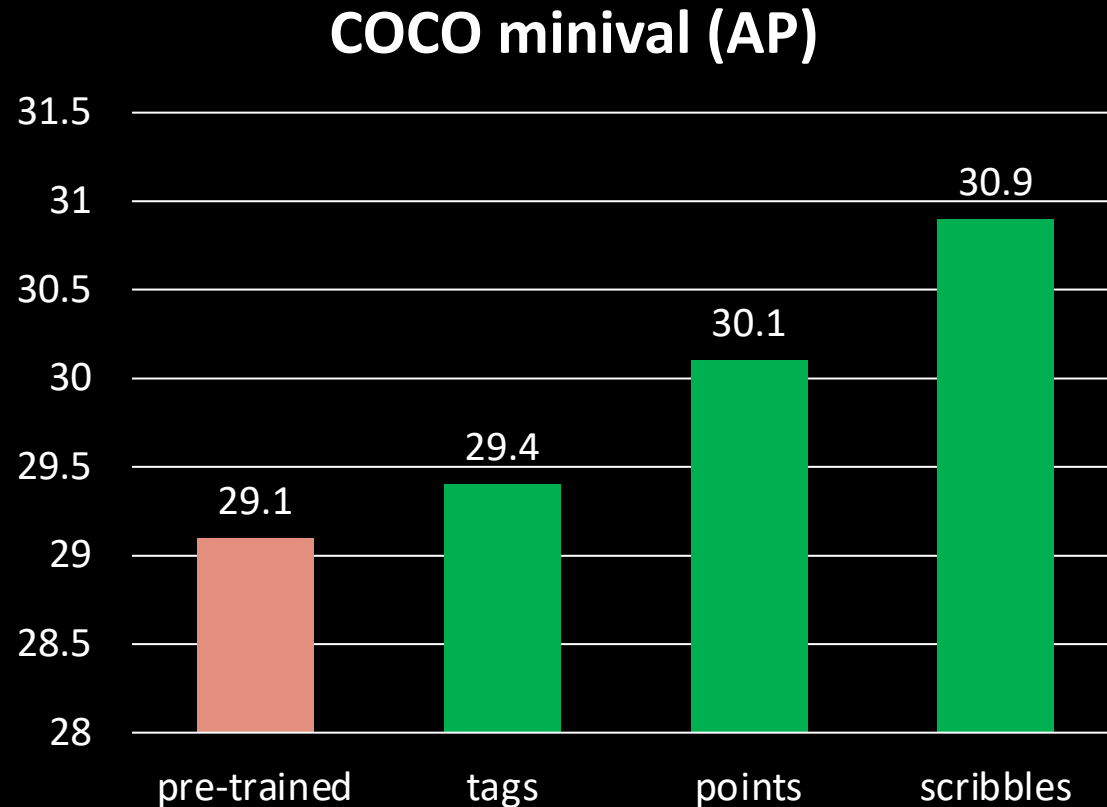
COCO val-2014 results (AP-50)



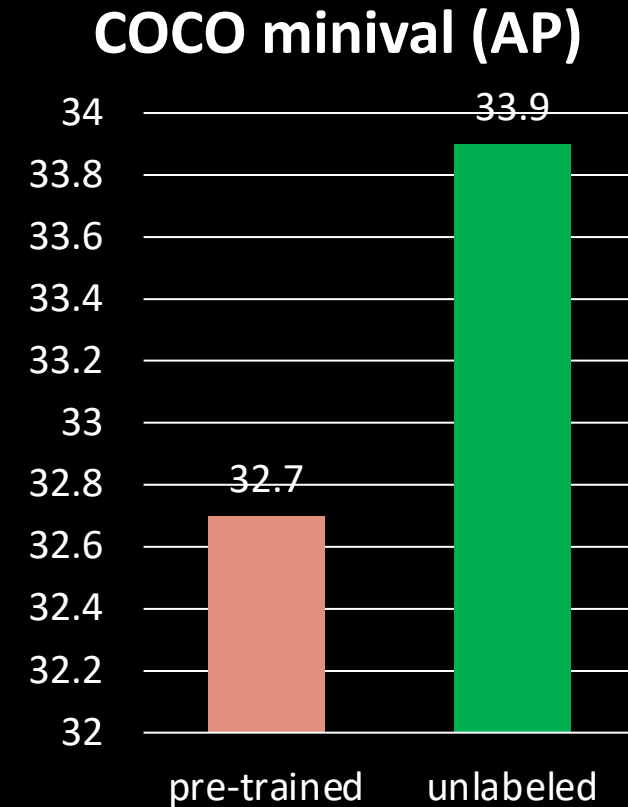
Training from scratch (single label)



# Experiments: Improve Pre-trained Models



**COCO-35**



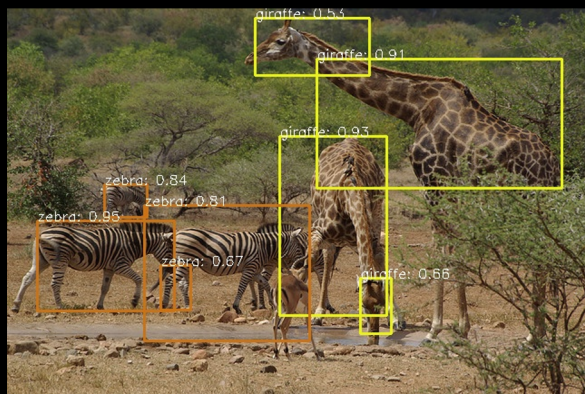
**COCO-115**

**Improving pre-trained models (Mixed labels)**

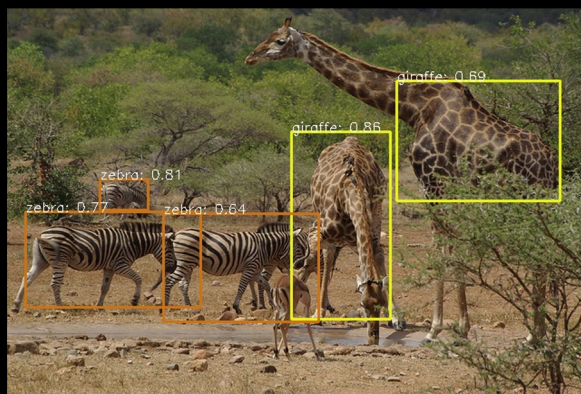


# Experiments

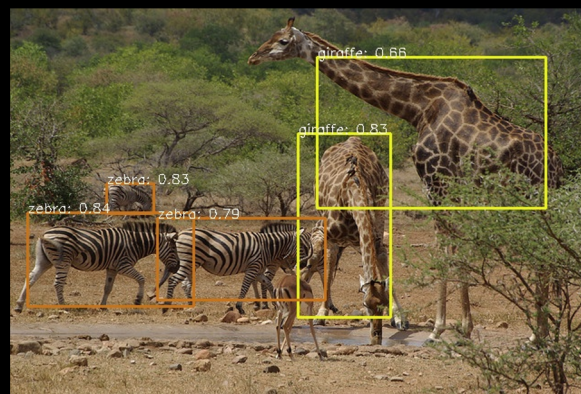
Tags



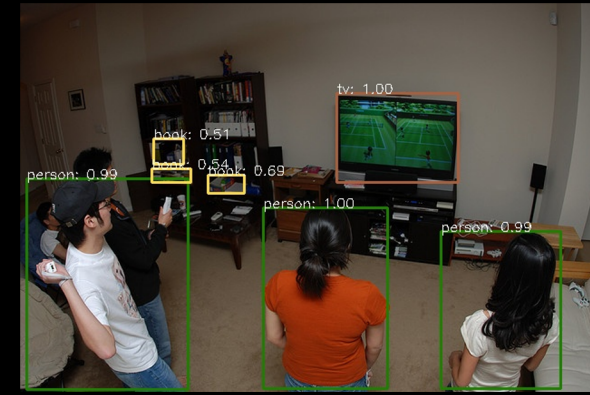
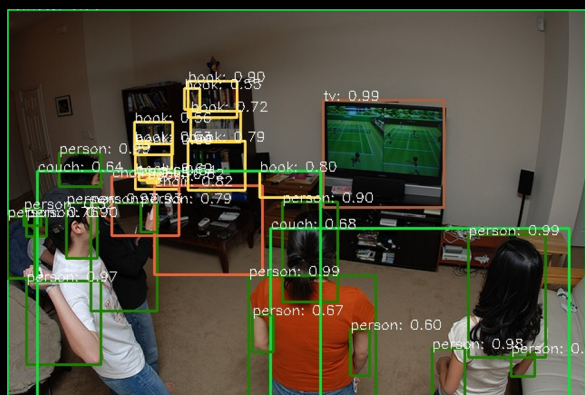
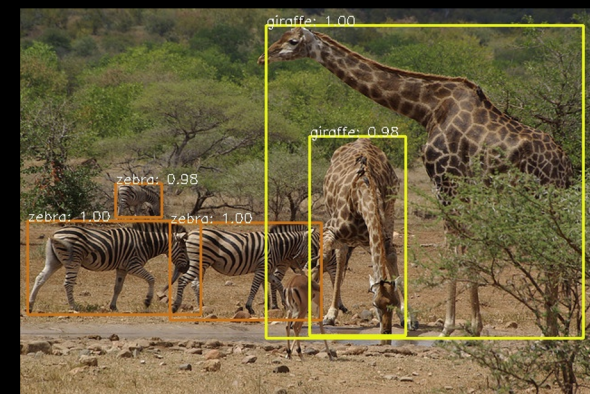
Points



Scribbles



Boxes



# Budget-aware Omni-supervised Detection

Approx. per-img budget on COCO:

- tags: 80s, points: 88.7s, scribbles: 160.4s, boxes: 346s

Given a fixed annotation budget (time), the most common strategy:

- **STRONG**: annotate using only boxes

UFO<sup>2</sup> allows a promising new policy:

- **N%B**: use N% budget for boxes and (1-N%) for points

Policy	Labels	# Labeled Images	AP
<b>STRONG</b>	<b>B + U</b>	<b>2312 + 7688</b>	<b>13.97 ± 0.98</b>
<b>80%B</b>	<b>P + B + U</b>	<b>1804 + 1850 + 6346</b>	<b>14.11 ± 1.01</b>



Thanks for watching!

