

MixTeacher: Mining Promising Labels with Mixed Scale Teacher for Semi-Supervised Object Detection

Liang Liu¹, Boshen Zhang¹, Jiangning Zhang¹, Wuhao Zhang¹, Zhenye Gan¹,
Guanzhong Tian³, Wenbing Zhu⁴, Yabiao Wang^{1*}, Chengjie Wang^{1,2*},

¹. Youtu Lab, Tencent ². Shanghai Jiao Tong University

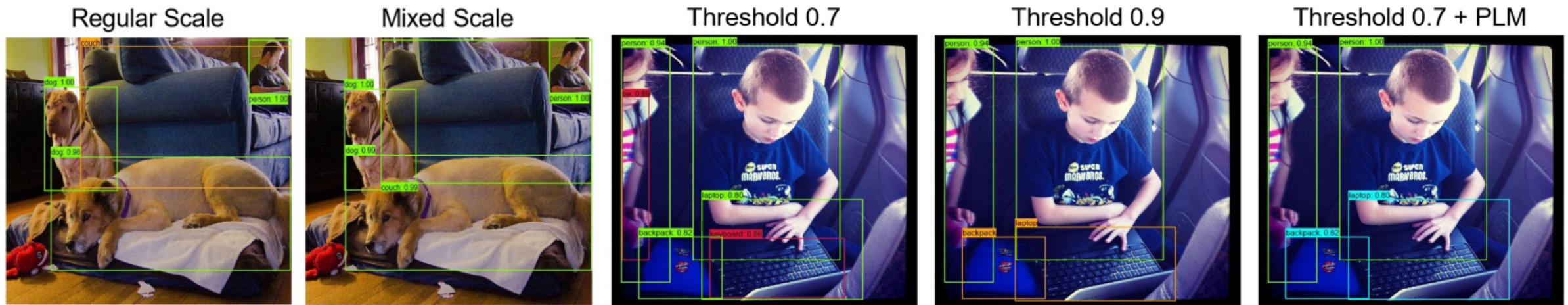
³. Ningbo Research Institute, Zhejiang University ⁴. Rongcheer Co., Ltd.

(* indicates Corresponding authors)

Poster: TUE-PM-310

MixTeacher Overview

- MixTeacher is a framework to address the challenge of scale variation in semi-supervised object detection.
- MixTeacher generates high-quality initial pseudo labels using a mixed-scale feature pyramid.
- MixTeacher mines low-confidence pseudo labels according to the score promotion of predictions across scales.
- MixTeacher achieves SoTA on various benchmarks under SSOD settings.



(a) Pseudo Labels Generation on Different Scales

(b) Pseudo Labels Filtering with Different Strategies

Background: Semi-Supervised Object Detection

Small amount of **labeled** images +
Large amount of **unlabeled** images

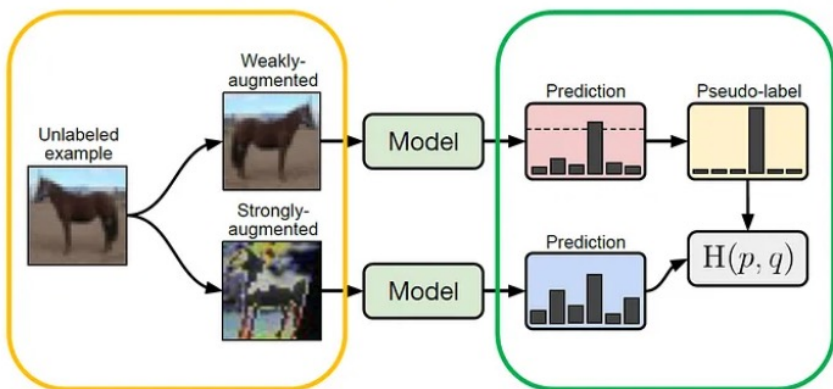


Semi-Supervised Learning

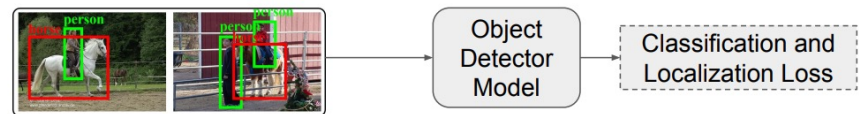
Semi-supervise in Image Classification

Input augmentations & perturbations

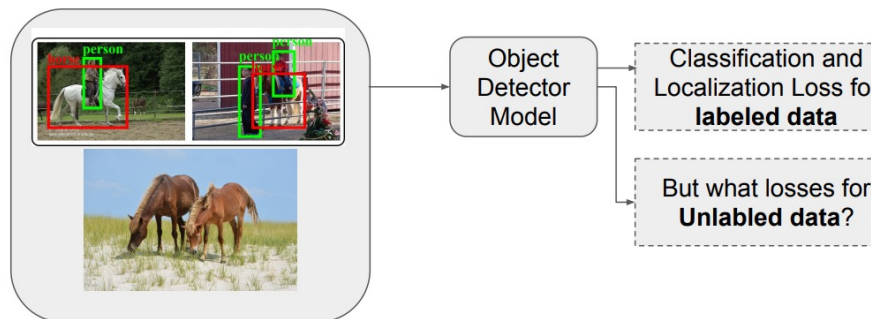
Consistency regularization



☺ Clear correspondence between the predictions in different views

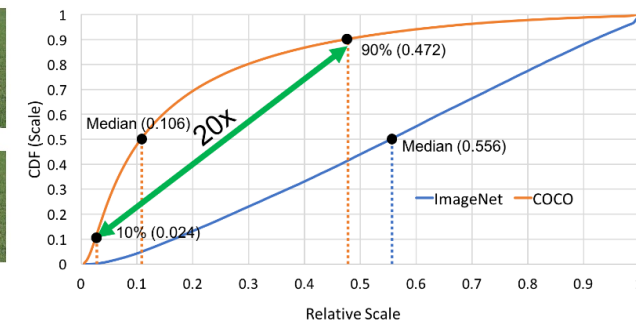
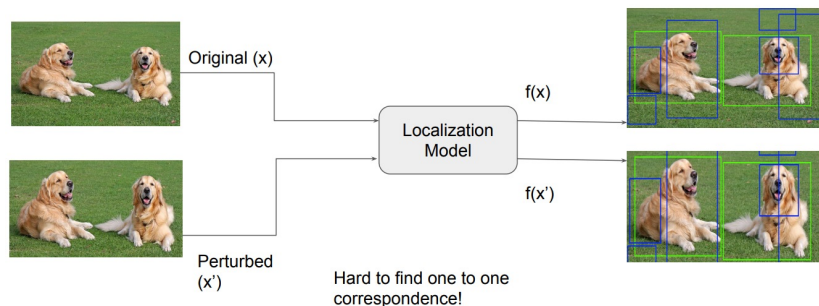


Supervised Learning



Semi-Supervised Learning

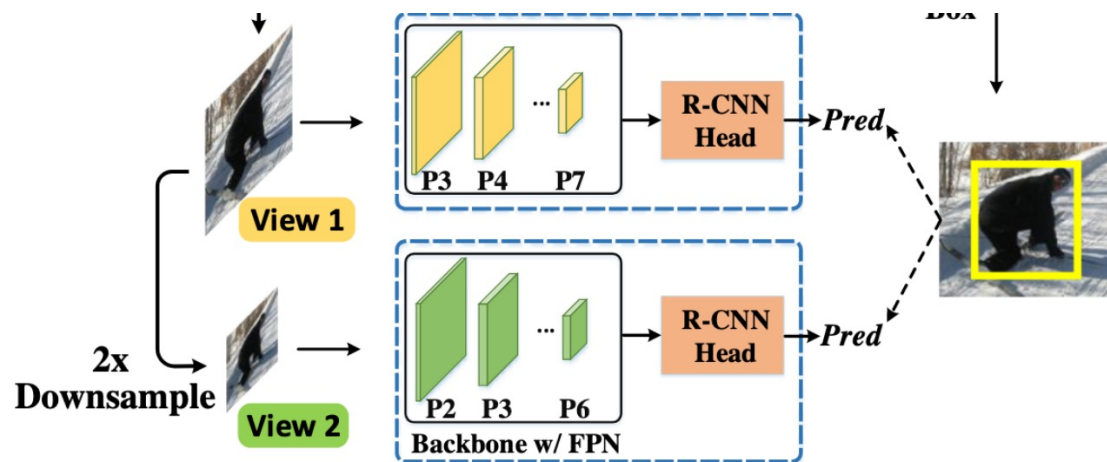
Semi-supervise in Object Detection



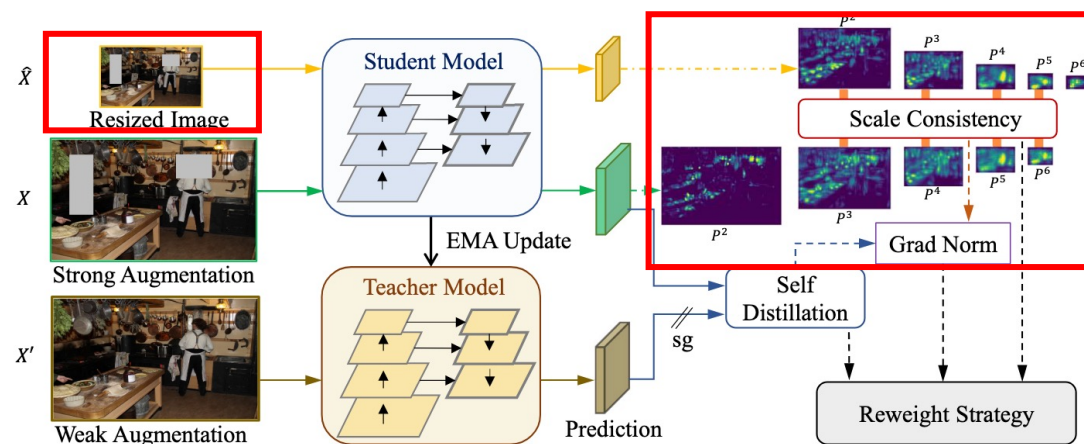
☹ Clear correspondence between the predictions in different views
☹ Object instance scale changes dramatically in detection task

Existing Observation: Additional Scales Helps SSOD

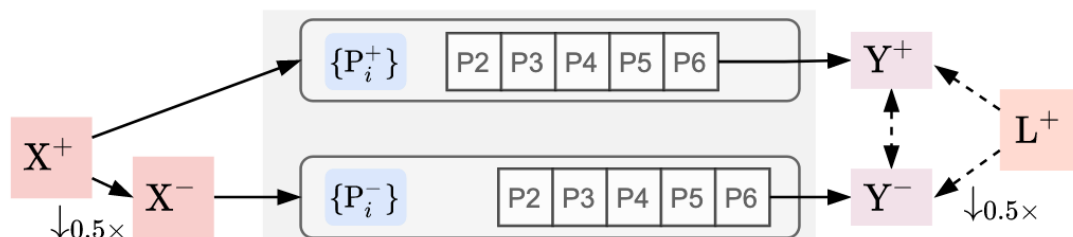
Additional **down-sampling scale** input helps SSOD



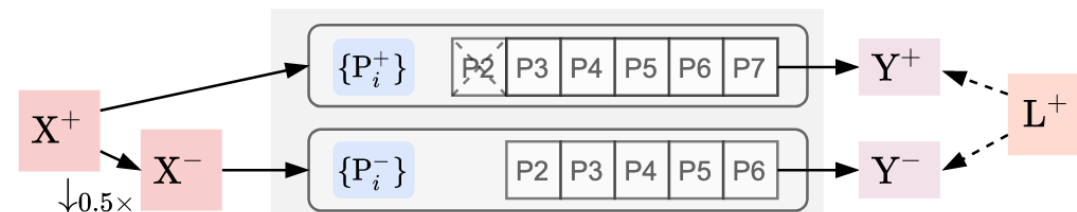
[PseCo, Li et al, ECCV 2022]



Unsupervised Branch
[SED, Guo et al, CVPR 2022]



(a) Scale Consistency Regularization in SED



(b) Multiview Scale-Invariant Learning in PseCo

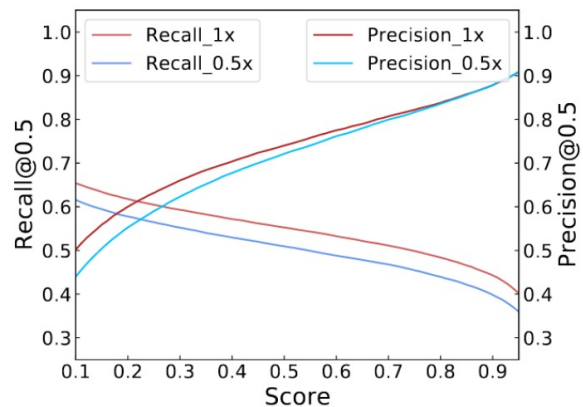
Previous methods adopt pseudo labels from **large scale inputs** and take additional scales as regularization.

Our Motivations: Regular Scale Pseudo Labels Are Not Always Reliable

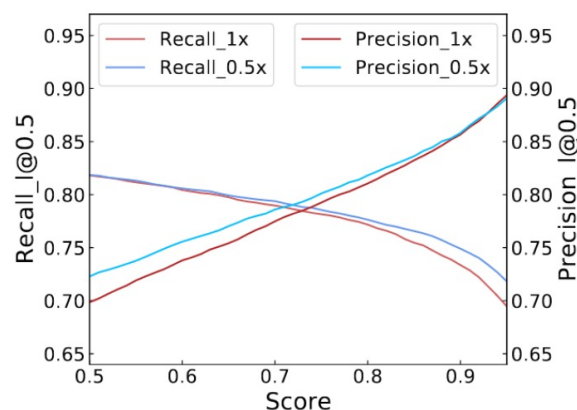
Input size significantly affects pseudo-labels in SSOD



Small scale input generates better predictions on large objects.



(a) Precision and recall for all objects



(b) Precision and recall for large objects

Model test on different scales

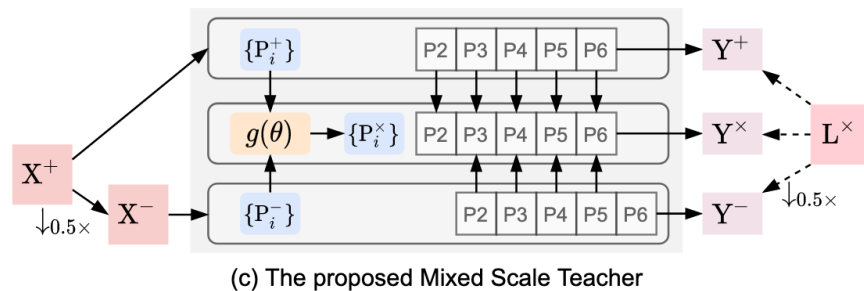
\mathbb{P}^+ : regular scale

\mathbb{P}^- : 0.5x down-sample scale

	mAP	AP_s	AP_m	AP_l
Test on \mathbb{P}^-	33.2	14.6	36.1	50.0
Test on \mathbb{P}^+	36.7	21.8	39.2	48.6

Method: Mixed Scale Teacher

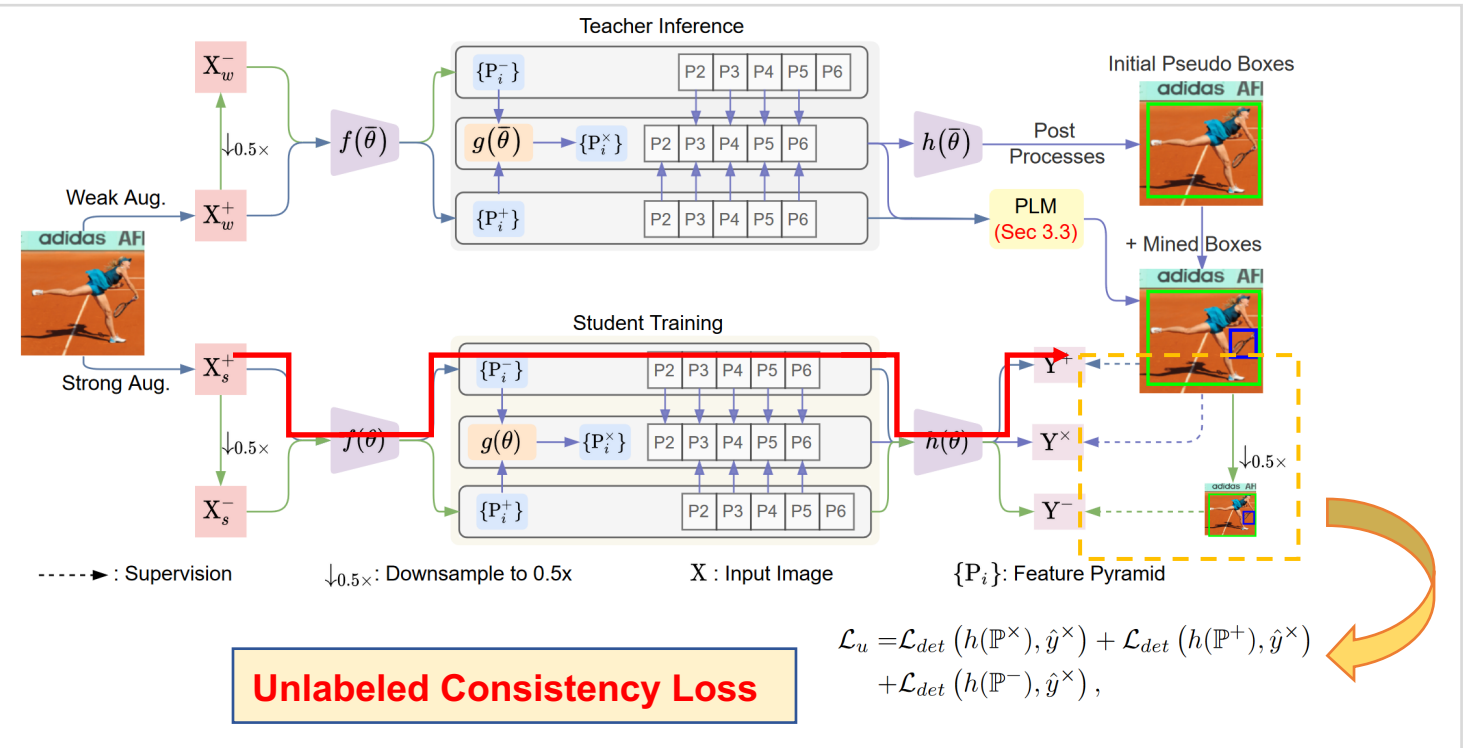
Pseudo Labels from Mixed Scale



$$\gamma = \sigma(g(P_i^+, P_{i-1}^- | \theta_g)),$$

$$P_i^\times = \gamma P_i^+ + (1 - \gamma) P_{i-1}^-.$$

Mixed Scale Feature Pyramid

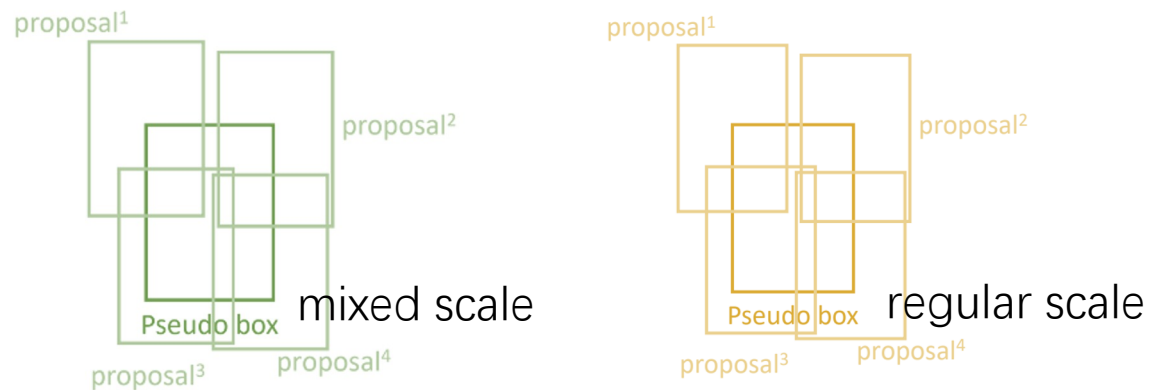


☺ **Efficient Training.** Only an additional module of two layer MLPs is insert into the original detector.

☺ **Fair Testing.** During testing, all parts not present in the original detector are discarded.

Method: Promising Labels Mining

Pseudo Labels from Lower Score Predictions :



Mean score promotion for a pseudo box across scales

$$\Delta q = \frac{1}{K} \left(\sum_{i=1}^K h_c(b_i, \mathbb{P}^\times | \bar{\theta}) - \sum_{i=1}^K h_c(b_i, \mathbb{P}^+ | \bar{\theta}) \right)$$

Procedure:

1. Construct a bag of proposals for each [low-confidence candidate](#).
2. Compute the mean score for mixed scale and regular scale.
3. Compute the gap for two scales as the mean score promotion.

Ablation Study:

Feature Scales			MST	PLM	mAP	AP ₅₀	AP ₇₅
\mathbb{P}^+	\mathbb{P}^-	\mathbb{P}^\times					
					26.8	45.1	28.4
✓					33.9 (+7.1)	54.0	37.0
✓	✓				34.7 (+7.9)	54.7	37.8
✓	✓	✓			34.4 (+7.4)	54.2	37.2
✓	✓	✓	✓		36.2 (+9.4)	56.5	39.5
✓	✓	✓	✓	✓	36.7 (+9.9)	57.0	39.7

Table 5. Analysis of various components of proposed approach. MST indicates generating pseudo labels from the mixed scale feature pyramid. PLM indicates promising labels mining.

Results of MixTeacher

Experimental comparison: Compared with 10+ methods on VOC 07, 12 and COCO 14 data sets, the best mAP has been achieved.

	COCO Partially Labeled				COCO Additional
	1%	2%	5%	10%	100 %
Supervised Baseline	12.15±0.27	16.65±0.18	21.45±0.16	27.10±0.07	40.9
STAC [35]	13.97±0.35	18.25±0.25	24.38±0.12	28.64±0.21	39.5 $\xrightarrow{-0.3}$ 39.2
SED [10]	COCO: Gains over baseline >10 mAP				40.2 $\xrightarrow{+3.2}$ 43.4
Unbiased Teacher [25]*	20.75±0.12	24.30±0.07	28.27±0.11	31.50±0.10	40.2 $\xrightarrow{+1.1}$ 41.3
Soft Teacher [40]†	20.46±0.39	-	30.74±0.08	34.04±0.14	40.9 $\xrightarrow{+3.6}$ 44.5
LabelMatching [3]*	25.81±0.28	-	<u>32.70±0.18</u>	35.49±0.17	40.3 $\xrightarrow{+5.0}$ 45.3
PseCo [17]†	22.43±0.36	27.77±0.18	32.50±0.08	36.06±0.24	41.0 $\xrightarrow{+5.1}$ 46.1
DTG-SSOD [16]†	21.27±0.12	26.84±0.25	31.90±0.08	35.92±0.26	40.9 $\xrightarrow{+4.8}$ <u>45.7</u>
Unbiased Teacher v2 [26]*	<u>25.40±0.36</u>	<u>28.37±0.03</u>	31.85±0.09	35.08±0.02	40.9 $\xrightarrow{+3.9}$ 44.8
MixTeacher (Ours)†	25.16±0.26	29.11±0.21	34.06±0.13	36.72±0.16	40.9 $\xrightarrow{+4.8}$ <u>45.7</u>

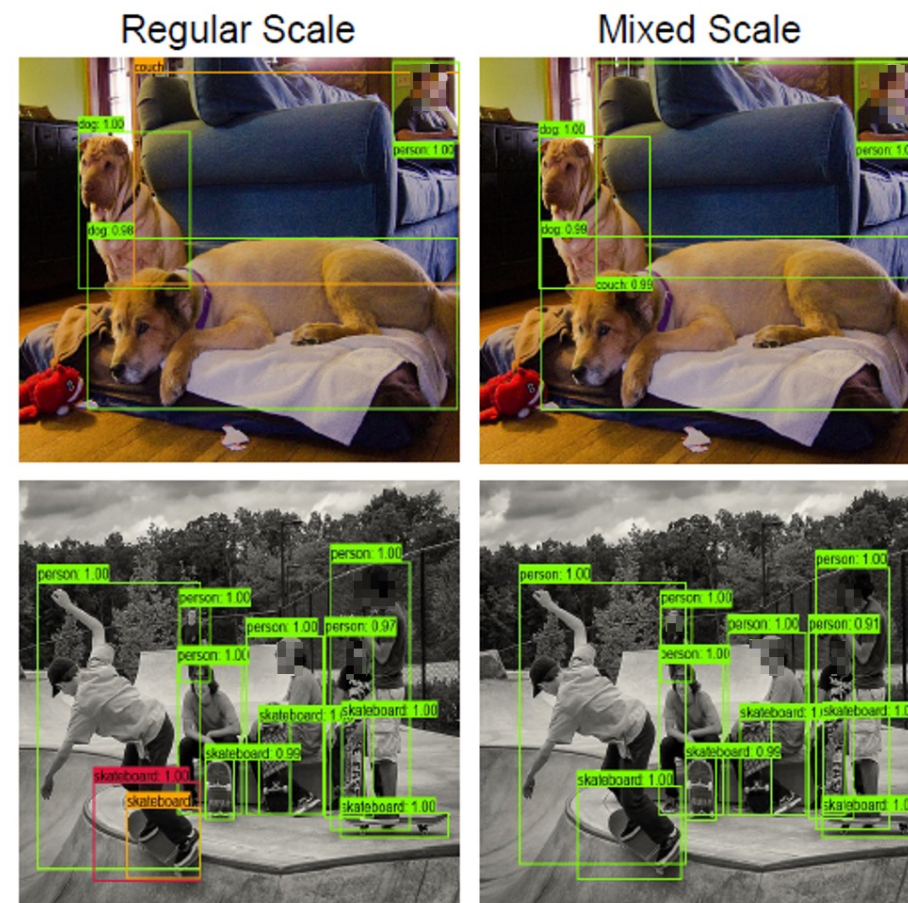
	AP ₅₀	AP _{50:95}
Supervised [25]	72.63	42.13
STAC [35]	77.45 (+4.82)	44.64 (+2.51)
Humble Teacher [36]	80.94 (+8.31)	53.04 (+10.91)
Rethinking Pse [18]	79.00 (+6.37)	54.00 (+11.87)
LabelMatching [3]	85.48 (+12.85)	55.11 (+12.98)
MixTeacher (Ours)	85.85 (+13.22)	56.25 (+14.12)

VOC: Gains over baseline > 13 mAP

Table 3. Experimental results on the VOC Additional setting.

	AP ₅₀	AP _{50:95}	Mixed Scale Teacher with FCOS			
			1%	2%	5%	10%
Supervised [25]	72.63	42.13				
STAC [35]	79.08 (+6.45)	46.01 (+3.88)				
Humble Teacher [36]	81.29 (+8.66)	54.41 (+12.28)				
Rethinking Pse [18]	79.60 (+6.79)	56.10 (+13.97)				
LabelMatching† [3]	85.81 (+13.18)	55.50 (+13.37)				
MixTeacher (Ours)	86.58 (+13.95)	56.83 (+14.70)				
Dense Teacher [46]			22.38	27.20	33.01	37.13
Unbiased Teacher v2 [26]			22.71	26.03	30.08	32.61
MixTeacher (Ours)			23.83	27.88	33.42	36.95

Table 4. Experimental results on the VOC Mixture setting.

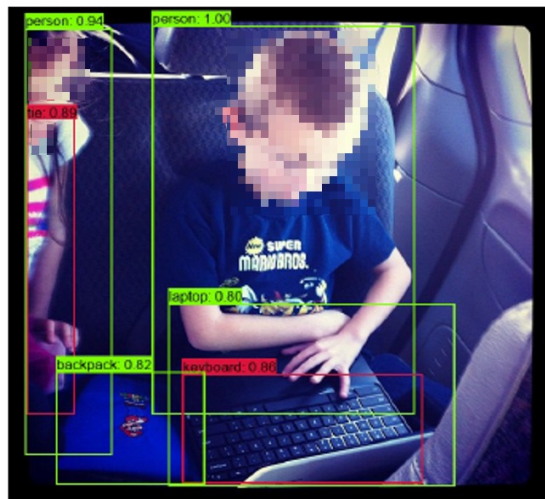


(a) Pseudo Labels Generation on Different Scales

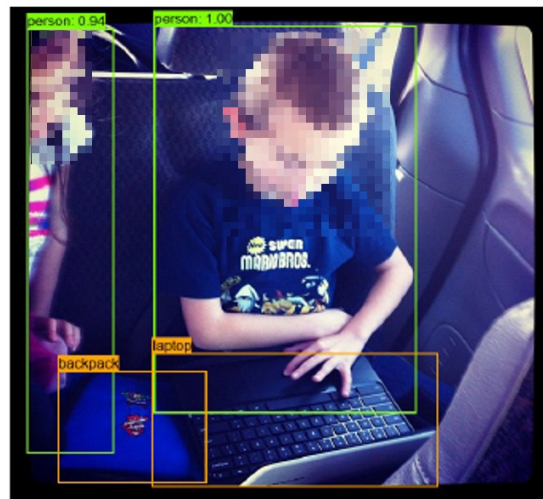
Pseudo-labels from mixed scales are more accurate than regular scales

More Results of MixTeacher:

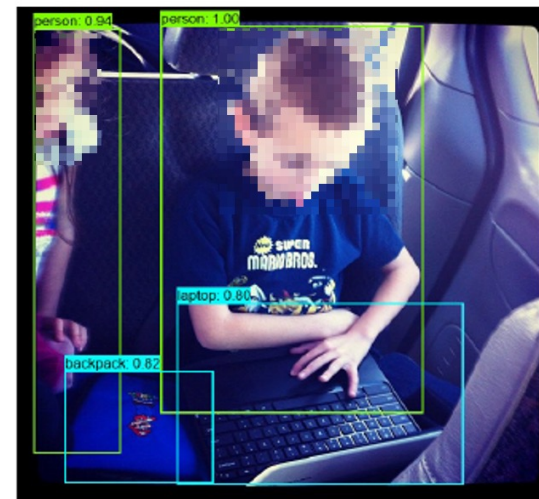
Threshold 0.7



Threshold 0.9

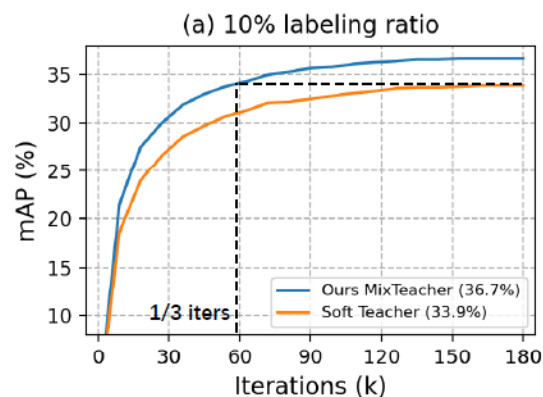


Threshold 0.7 + PLM

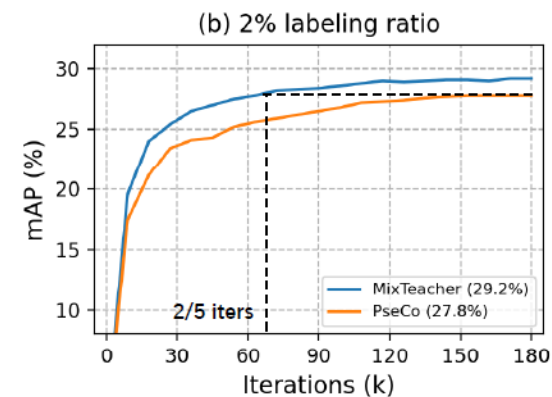


Pseudo Labels Visualization.

Green Boxes: TP, Red Boxes: FP, Orange Boxes: FP, Blue Boxes: Correct Mined Boxes



Achieves **Baseline** Performance with **1/3** Iterations



Achieves **SOTA** Performance with **2/5** Iterations

MixTeacher: Mining Promising Labels with Mixed Scale Teacher for Semi-Supervised Object Detection

Thanks

Poster: TUE-PM-310