

Supplementary: Few-Shot Domain Adaptive Object Detection for Microscopic Images

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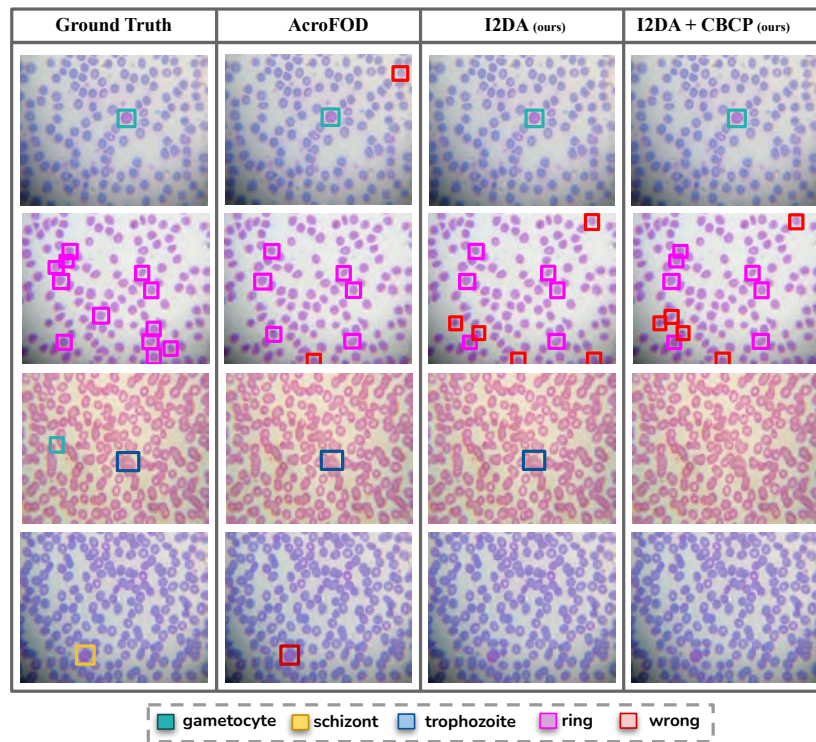


Fig. 1: Qualitative results of failed examples of LCM Malarial defected regions after adaptation. From Left to right, Ground Truth, AcroFOD, I2DA(ours) without augmentation, I2DA+cbcp(ours) with augmentation.

YOLOv5 return best.pt and last.pt weights of the trained model as per the validation data. Paper contains the last.pt results.

Table 1: Malaria test set - mAP@50(%) on best.pt.

Malaria-HCM-1000x → Malaria-LCM-1000x															
Method	mAP@50(%)			Gametocyte			Schizont			Trophozoit			Ring		
Shots	2	3	5	2	3	5	2	3	5	2	3	5	2	3	5
AsyFOD	26.0	28.0	29.7	14.9	17.1	23.0	1.2	0.8	1.8	59.4	62.5	63.6	28.7	31.8	30.3
AcroFOD	27.2	<u>28.7</u>	<u>40.4</u>	16.6	<u>22.7</u>	<u>60.5</u>	0.9	1.1	<u>7.8</u>	<u>62.5</u>	59.2	62.1	29.1	<u>31.8</u>	<u>31.2</u>
Ours	38.7	38.8	48.9	57.6	54.2	68.2	2.4	6.6	30.4	65.7	<u>61.7</u>	65.7	<u>28.9</u>	32.6	31.5

Table 2: Raabin-WBC test set - mAP@50(%) on best.pt.

Raabin-WBC-HCM → Raabin-WBC-LCM															
Method	mAP50(%)			Large Lymph			Neutrophil			Small Lymph			Monocyte		
Shots	2	3	5	2	3	5	2	3	5	2	3	5	2	3	5
AsyFOD	36.2	38.7	32.2	41.0	45.8	49.8	62.6	53.3	56.4	40.7	45.2	22.1	0.3	10.4	0.3
AcroFOD	<u>54.5</u>	<u>49.2</u>	<u>64.5</u>	75.0	<u>63.0</u>	<u>83.0</u>	92.3	<u>88.9</u>	94.3	34.1	<u>38.6</u>	66.2	<u>16.7</u>	<u>6.30</u>	<u>14.4</u>
Ours	<u>55.2</u>	<u>58.8</u>	<u>70.7</u>	71.9	68.0	86.2	<u>80.1</u>	91.0	<u>93.8</u>	<u>37.3</u>	37.1	<u>44.2</u>	31.6	39.0	58.7

Table 3: best.pt results in mAP@50(%) on M5 and RWBC test sets on 8 random few-target images.

Data	Malaria					Raabin-WBC				
	mAP@50	Gamet.	Schizo.	Troph.	Ring	mAP@50	L-Lymp.	Neutro.	S-Lymp.	Mono.
AsyFOD	26.2	4.5	6.8	62.3	31.4	30.2	28.9	50.5	40.3	0.9
AcroFOD	34.7	50.3	<u>5.4</u>	60.6	22.5	50.8	59.1	59.5	58.1	<u>26.3</u>
Ours	42.2	68.8	3.0	65.8	<u>31.3</u>	55.7	<u>63.5</u>	<u>79.6</u>	45.0	34.5

Algorithm 1 Get Increment Stats

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Input:  $\mathcal{O}_c, \mathcal{I}_f$ 
 $inc_{stats} \leftarrow \{\}$ 
 $max \leftarrow \max(\mathcal{O}_c)$ 
for each  $c$  in keys of  $\mathcal{O}_c$  do
  if  $max > \mathcal{O}_c[O]$  then
     $S^c \leftarrow \frac{max - \mathcal{O}_c[c]}{\mathcal{O}_c[c]}$ 
     $total_{i_f} \leftarrow \frac{S^c \times \mathcal{O}_c[c]}{\text{len}(\mathcal{I}_f)}$ 
    if  $total_{i_f} < 0$  then
       $total_{i_f} \leftarrow 1$ 
    end if
     $times_{i_f} \leftarrow \frac{S^c}{total_{i_f}}$ 
     $inc_{stats}[c] \leftarrow [\mathcal{O}_c[c], times_{i_f}, total_{i_f}]$ 
  end if
end for
return  $inc_{stats}$ 

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