

Efficient Precision Control in Object Detection Models for Enhanced and Reliable Ovarian Follicle Counting - Supplementary material

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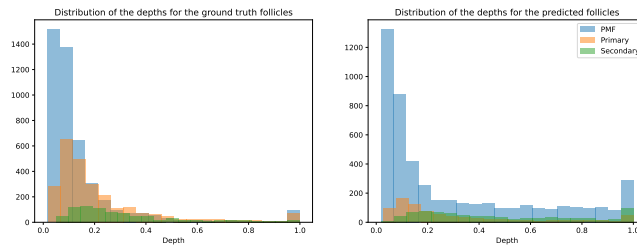


Fig. 1: Distribution of the depths of the ground truth and predicted follicles

Algorithm 1: Associate depth to each predicted bounding box

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Data:  $\mathcal{D} = \{X_1, \dots, X_n\}$ 
foreach element  $x$  in  $\mathcal{D}$  do
   $B \leftarrow \hat{f}(x)$ ;
   $C \leftarrow \text{FindContour}(x)$  ▷ Find the contour of the ovary;
  foreach box  $b \in B$  do
     $C_{dilated} \leftarrow C$ ;
    while  $\text{Center}(b) \notin C_{dilated}$  do
       $C_{dilated} \leftarrow \text{DilateContour}(C_{dilated})$  ▷ Dilate the contour of the ▷ ovary until
      the box is inside
    end
     $depth \leftarrow \frac{\text{Area}(C_{dilated})}{\text{Area}(x)}$  ▷  $\text{Area}(x)$  is the area of the ovary  $b \leftarrow \text{AddToList}(b, depth)$ 
    ▷ Add depth to box information  $\text{Update}(B, b)$  ▷ Update the predictions with the
    new information
  end
end

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Table 1: Training Parameters of the Object Detectors and Classification models

EfficientDet Training Parameters		Yolo Training Parameters	
Parameter	Value	Parameter	Value
Architecture	EfficientDet-d2	Architectures	Yolov8-Large
Learning Rate	0.001	Learning Rate	0.01
Epochs	100	Epochs	100
Batch Size	25	Batch Size	16
Image Size	1000	Image Size	640
Prediction Confidence Threshold	0.3	Pretrained	True
WBF IoU Threshold	0.3	Optimizer	Adam
Number of Workers	4	Mask Ratio	4
LR Warmup Epoch	1	IoU Threshold	0.7
		Max Detections	300

VGG16 Training Parameters	
Parameter	Value
Architectures	VGG16
Learning Rate	10^{-6}
Weight Decay	0.01
Epochs	24
Batch Size	16
Image Size	512
Pretrained	True

Table 2: Precision and Recall for EfficientDet and Yolo models

Class	EfficientDet			YOLO		
	Precision (%)	Recall (%)	mAP (%)	Precision (%)	Recall (%)	mAP (%)
All Classes	29.8	83.8	32.8	44.2	74.9	33.7
Primordial	24.9	88.1		42.1	71.9	
Primary	58.3	87.0		58.3	82.6	
Secondary	23.4	96.4		44.2	74.9	