



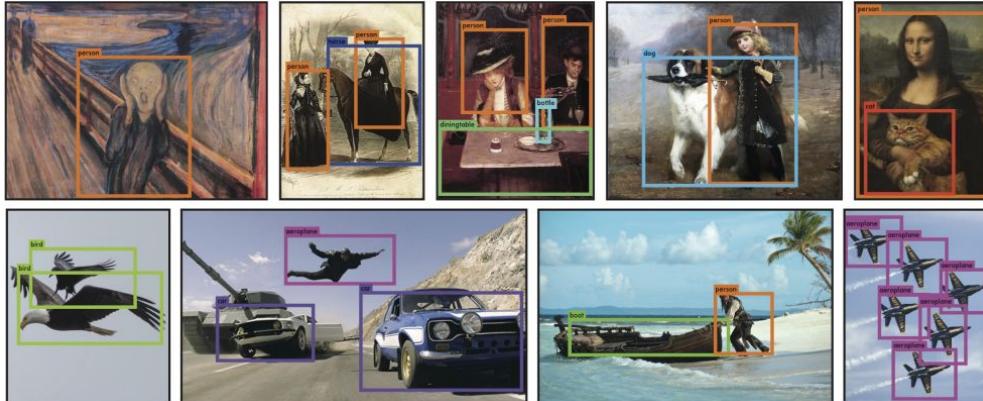
Supervised Team YOLOv8 sub-team

Arthur Caillau, Icxa Khandelwal,
Sumit Sakarkar, Julieta Millan

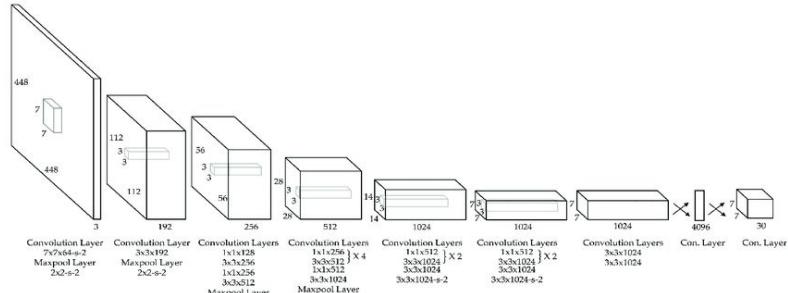


YOLO

Real-time object detection algorithm



24 convolutional layers followed by 2 fully connected layers, pretrained on ImageNet



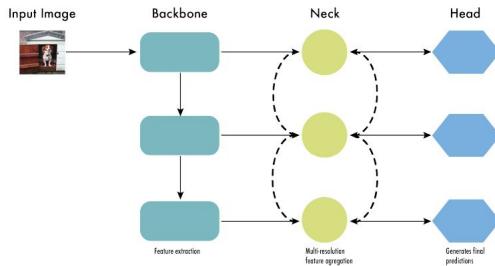
You Only Look Once: Unified, Real-Time Object Detection (Redmon, 2015)



YOLOv8 (seg)

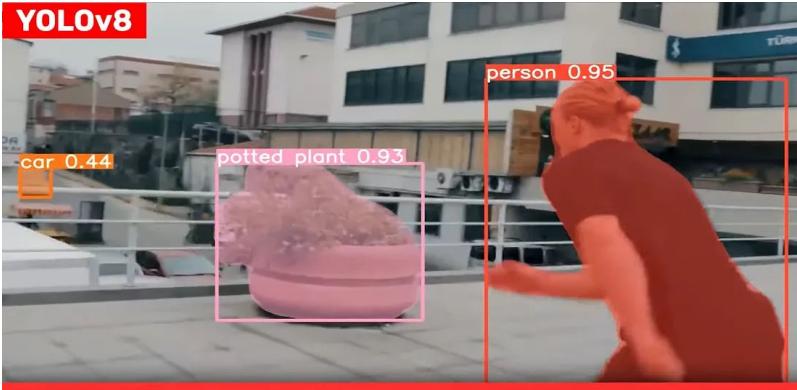
Latest implementation of YOLO, by Ultralytics (2023)

- Advanced Backbone and Neck Architectures
- Pretrained on [COCO](#)



A Comprehensive Review of YOLO: From YOLOv1 and Beyond (Terven, 2023)

YOLOv8



YOLOv8 Object Detection + Instance Segmentation

Instance Segmentation: **identifying**, **separating** and **classifying** individual objects within an image. The output is a set of **masks or contours**, **class labels** and **confidence scores** for each object



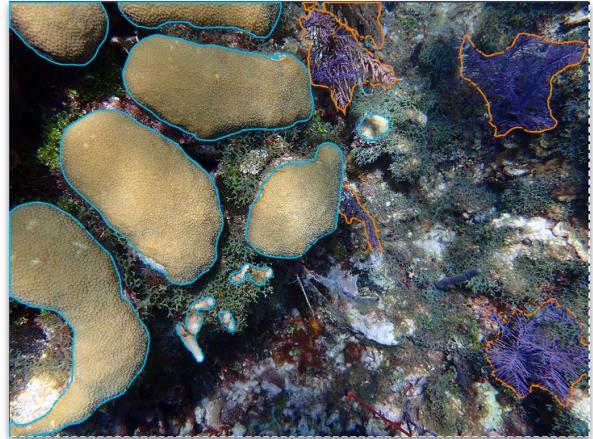
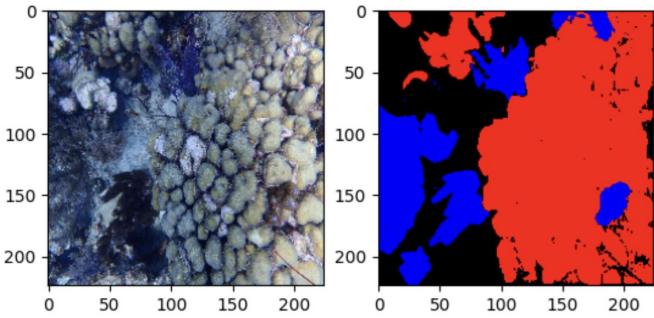
Methodology

Using the benthic dataset, labels and masks (rs_labelled):

- Transform the **image segmentation masks** into YOLOv8 **PyTorch TXT** format ([Notebook 1](#))
- Transform **files and folder structures** into format directly usable by YOLOv8 ([Notebook 2](#))
- Train the baseline models ([Notebook 3](#) and [4](#))

```
0 0.443359375 0.7646484375 0.443115234375 0.7  
1 0.826171875 0.62646484375 0.826171875 0.626  
1 0.06475830078125 0.58251953125 0.0645141601  
0 0.37744140625 0.2396240234375 0.37719726562  
0 0.31005859375 0.8251953125 0.309814453125 0  
1 0.001750946044921875 0.9013671875 0.0015010  
0 0.736328125 0.489990234375 0.736328125 0.49  
0 0.163818359375 0.9453125 0.16357421875 0.94
```

Yolov8 PyTorch txt format





Preliminary results

Two baseline models

- Region specific (trained on SEAFLOWER_BOLIVAR)
- All regions

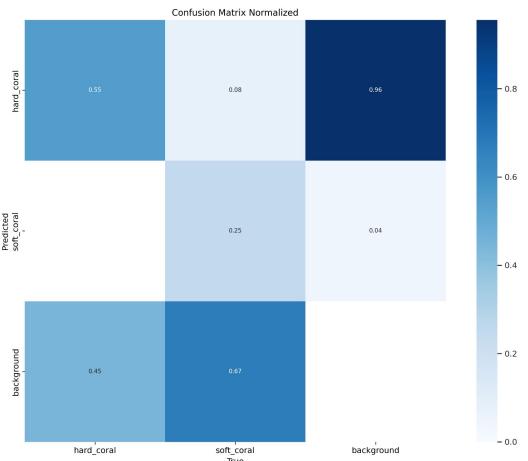
Test what works best for these models and datasets -
Specific models might achieve better performance at the
expense of worse generalization



Baseline 1

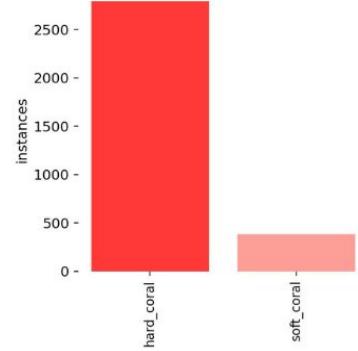
Data from only one region - SEAVIEW_BOLIVAR

- Task: instance segmentation
- Pretrained model size: m (~27M parameters)
- Number of epochs: 20 (~ training time: 20min)
- Dataset: SEAFLOWER_BOLIVAR, 80% random split train/val

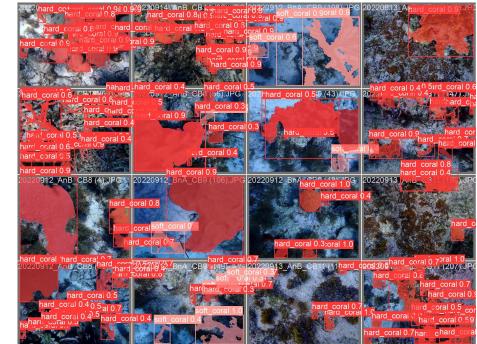


Performance

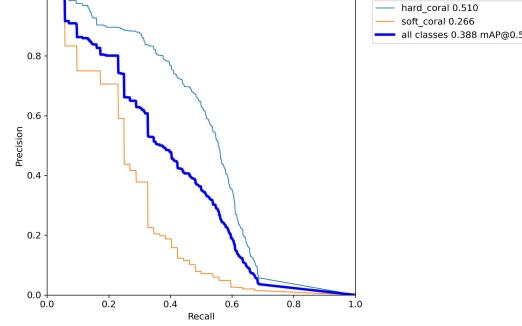
Labels



Predictions



Precision-Recall Curve

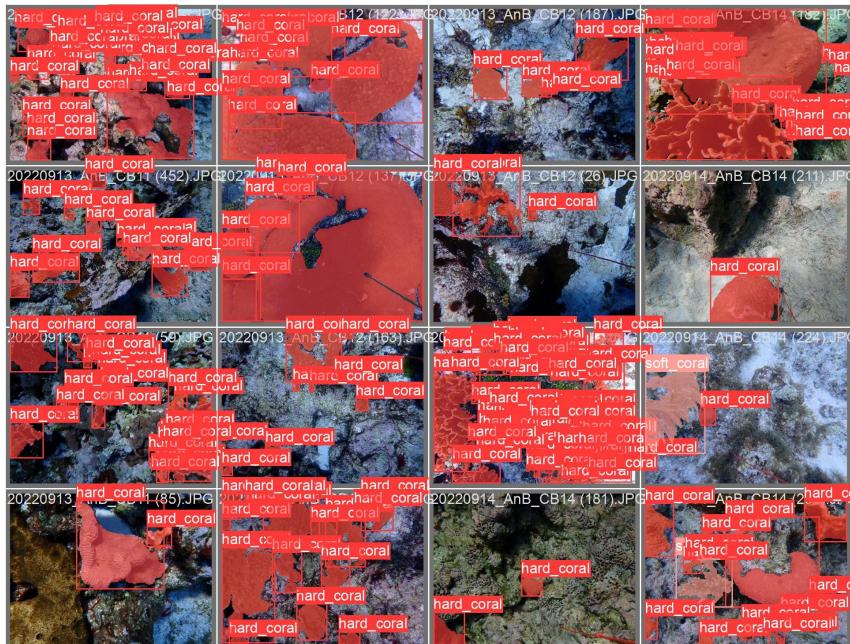




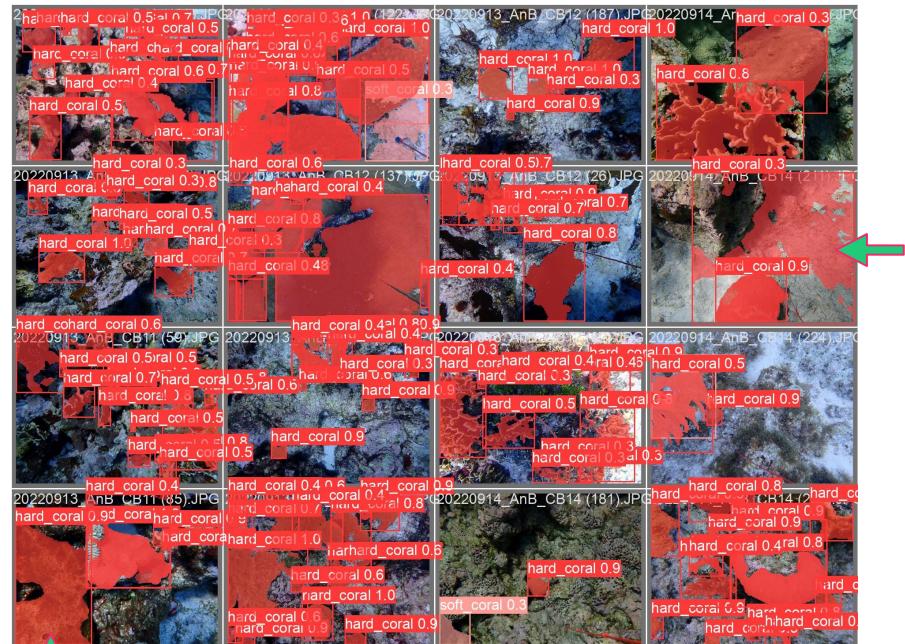
Baseline 1

Data from only one region - SEAVIEW_BOLIVAR

Ground truth



Prediction

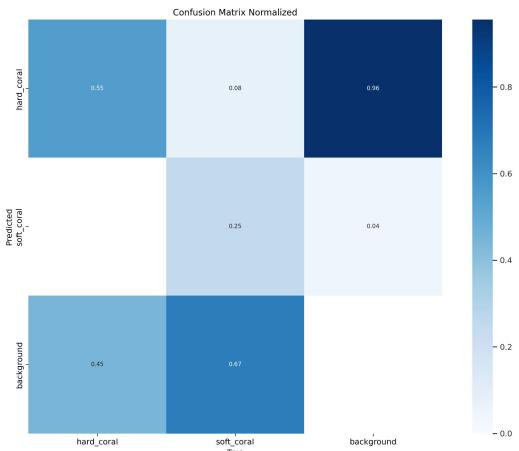




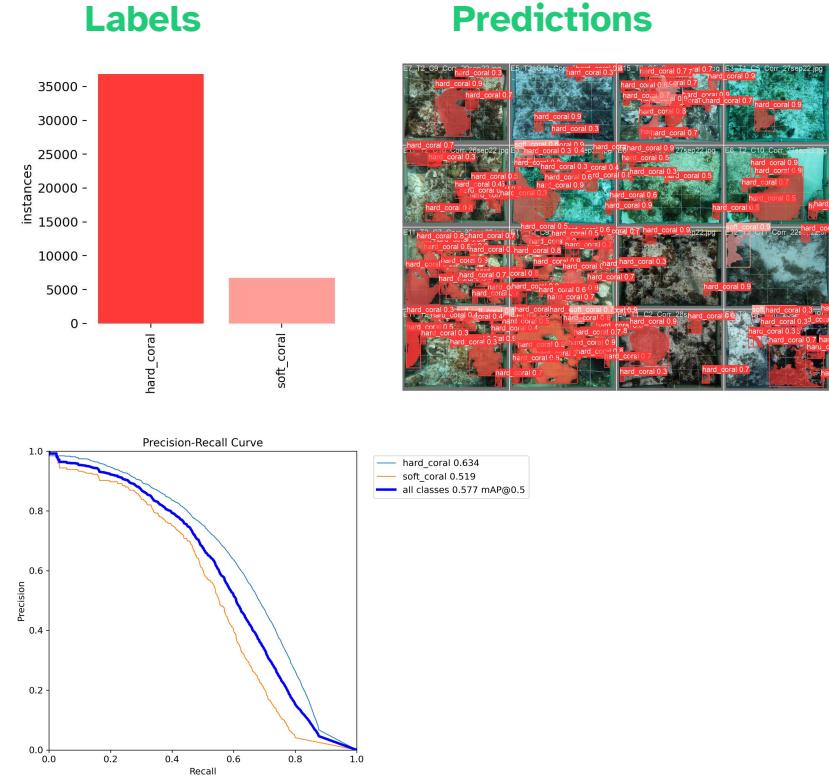
Baseline 2

Data from all regions (but SEAVIEW_PAC_AUS)

- Task: instance segmentation
- Pretrained model size: m (~27M parameters)
- Number of epochs: 20 (~ training time: 1h40m)
- Dataset: All regions except SEAVIEW_AUS - 80% random split train/val



Performance

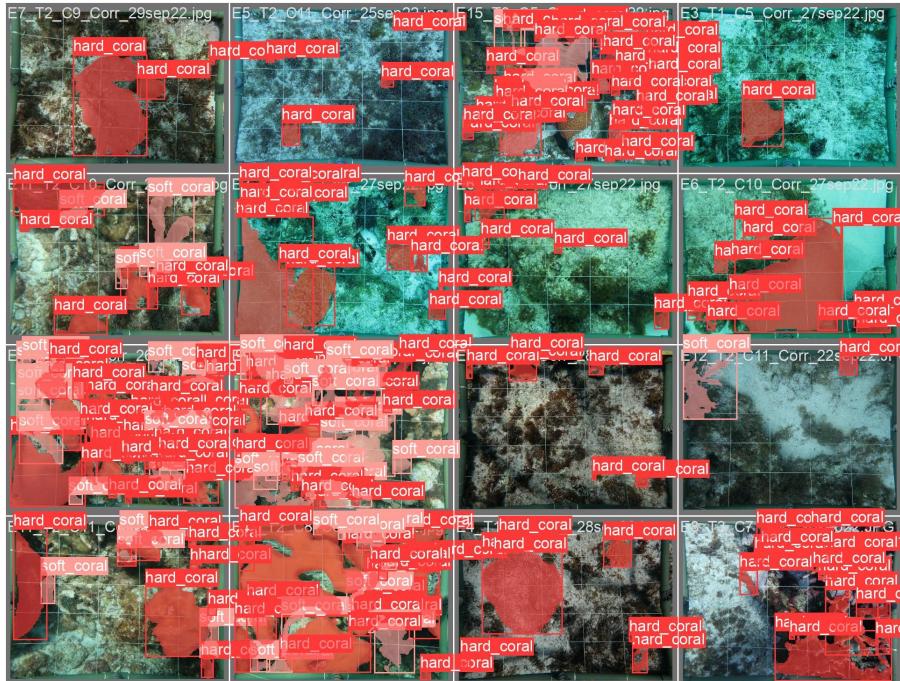




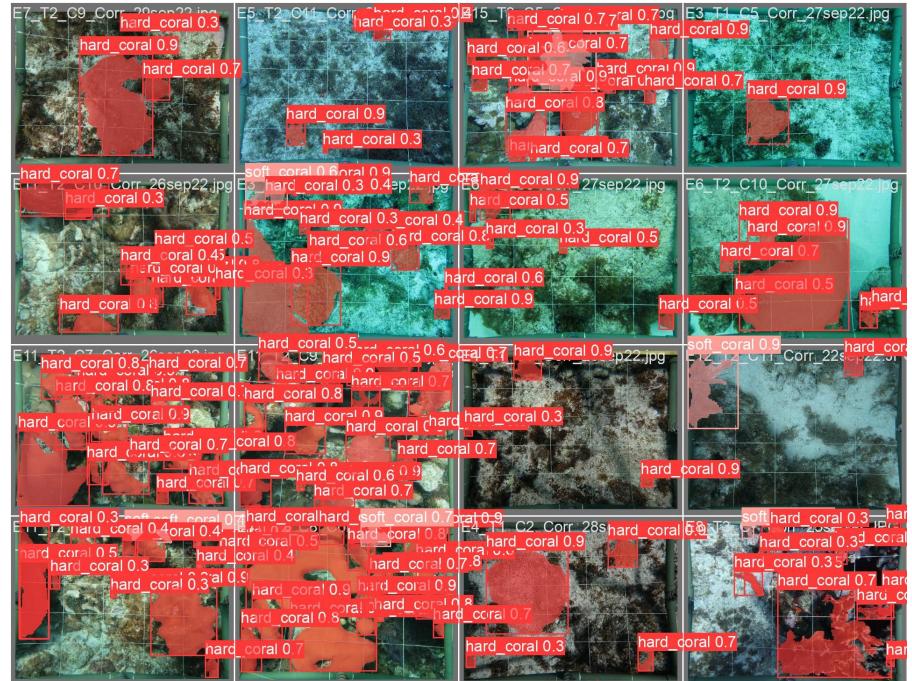
Baseline 2

Data from all regions (but SEAVIEW_PAC_AUS)

Ground truth

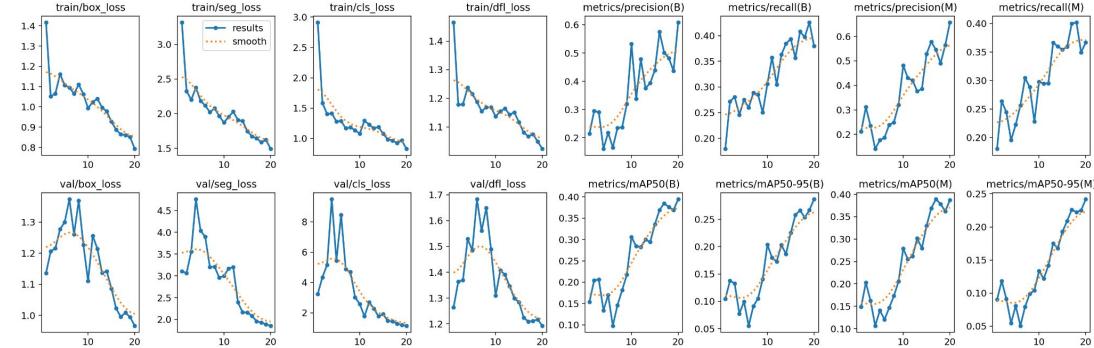


Prediction



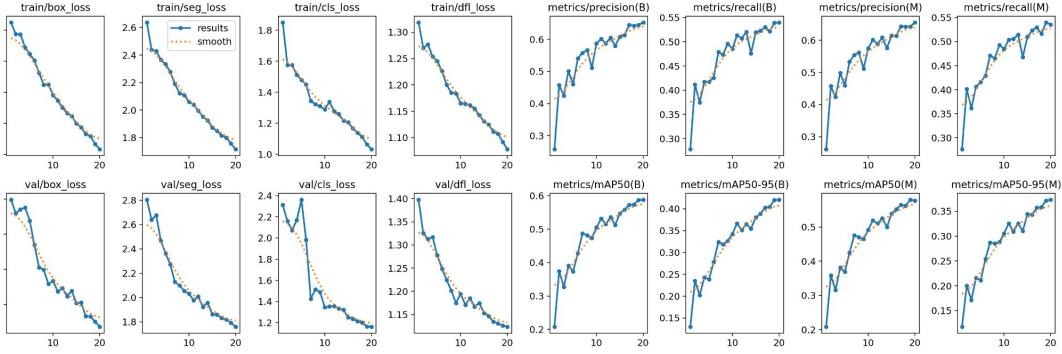


The baselines can still learn a lot!



Baseline 1 - region specific

Baseline 2 - All regions



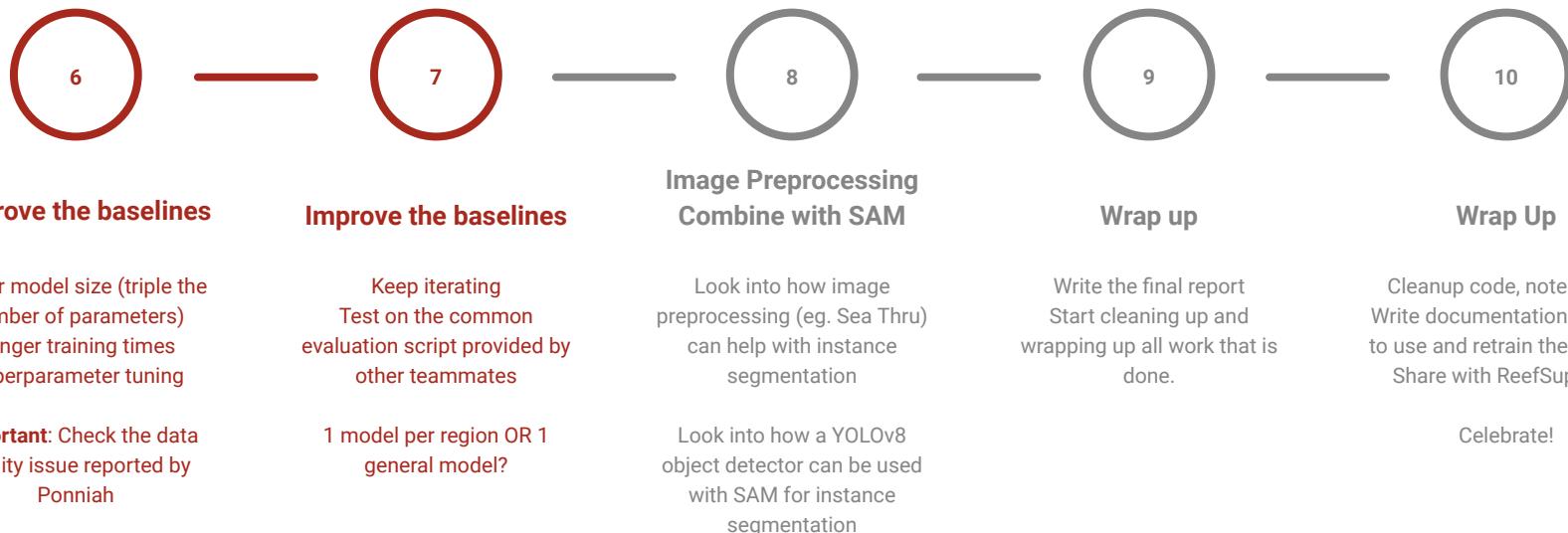


Conclusions and next steps

- YOLOv8 seems like a good model for instance segmentation and can be trained on the available dataset.
- Higher **computer power** to train would still improve performance
- **Bigger YOLOv8 models** should be tried to better fit the data - we are currently using the middle size model `m` but large `l` and xlarge `x` are also available.
- It would be great to generate a **confusion matrix that reports pixel classification** errors. That would provide an idea on how the model will be used downstream (to evaluate **benthic coverage**)
- Consider other metrics



Roadmap





FruitPunch AI

Questions?