

Object Detection on Digital Breast Tomosynthesis Images with YOLO

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Background

- Digital Breast Tomosynthesis (DBT)
 - More accurate
 - More time-consuming
- Automated methods
 - Computer-aided detection
 - Computer-aided diagnosis
- Objective
 - Object detection on DBT
 - Lower specificity and precision (while not resulting in a significant increase in sensitivity)



Method: You Only Look Once (YOLO)

- Object detection
 - Finding all the objects in an image and drawing the so-called bounding boxes around them
 - Two general groups
 - Algorithms based on classification: RCNN, Fast-RCNN, RetinaNet, etc.
 - Algorithms based on regression: YOLO, Single Shot Multibox Detector (SSD), etc.
- Regression-based object detection
 - Faster, but less accurate in general
- YOLO
 - Reframe object detection as **a single regression problem (convolution network)**
 - **You Only Look Once** at an image to predict what objects are present and where they are
 - Uses features from the **entire image** to predict
 - Predicts **all bounding boxes** across **all classes** for an image **simultaneously**

Method: You Only Look Once (YOLO)

- Unified detection

- Divides the input image into an $S \times S$ grid
- Each grid cell predicts B bounding boxes and confidence scores for those boxes

$$\Pr(\text{Object}) * \text{IOU}_{\text{pred}}^{\text{truth}}$$

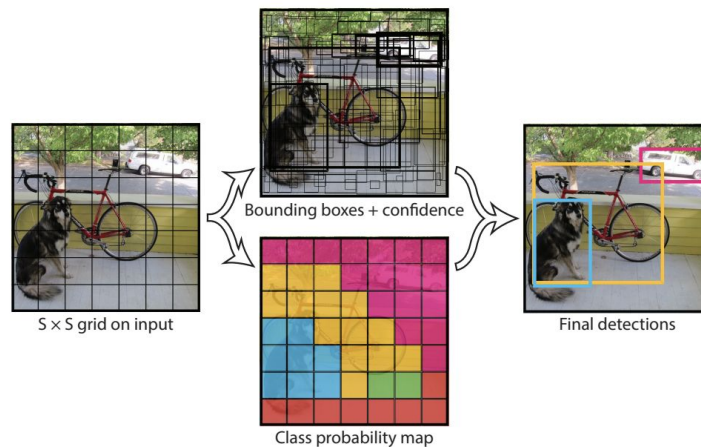
(IOU: intersection over union)

- Each bounding box consists of 5 predictions
 - (x, y) : center of the box
 - **width, height**
 - **confidence**
- Each grid cell also predicts C conditional class probabilities

$$\Pr(\text{Class}_i | \text{Object})$$

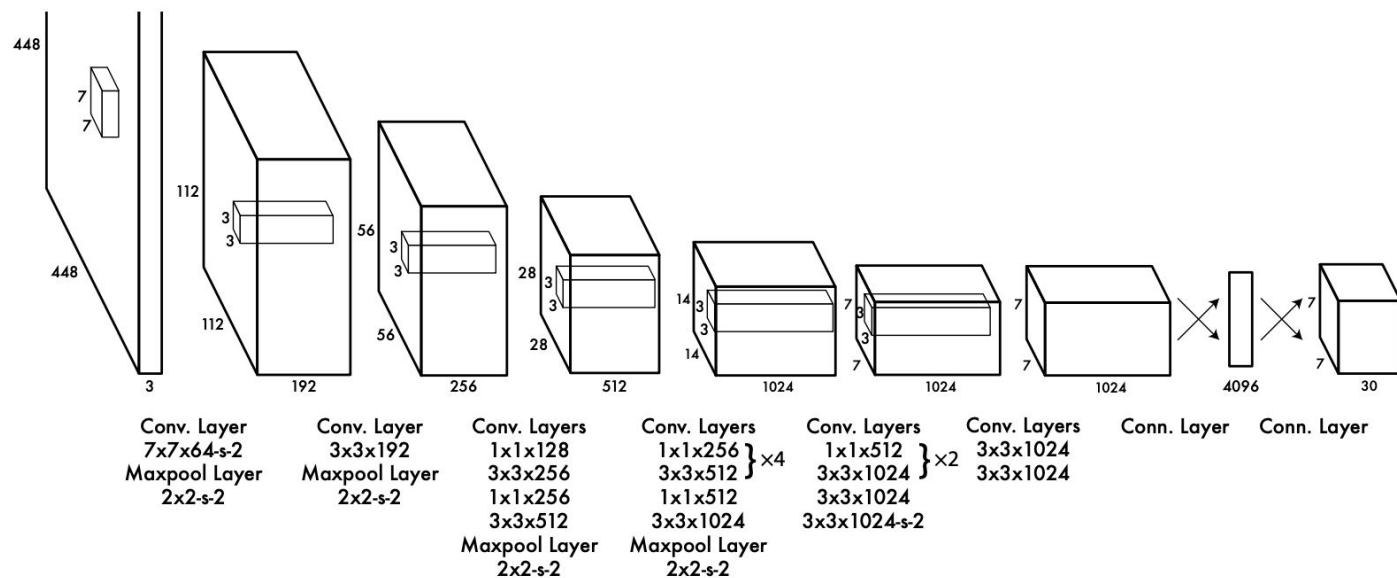
- At test time: class-specific confidence scores for each box

$$\Pr(\text{Class}_i | \text{Object}) * \Pr(\text{Object}) * \text{IOU}_{\text{pred}}^{\text{truth}} = \Pr(\text{Class}_i) * \text{IOU}_{\text{pred}}^{\text{truth}}$$



Method: You Only Look Once (YOLO)

- Full network architecture
 - 24 convolutional layers followed by 2 fully connected layers
 - 1×1 reduction layers
 - Optimize for sum-squared error



Method: You Only Look Once (YOLO)

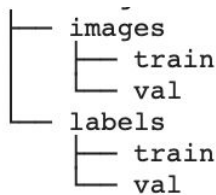
- Pros

- Extremely fast
- Reasons globally about the image when making predictions
- Learns generalizable representations of objects

- Cons

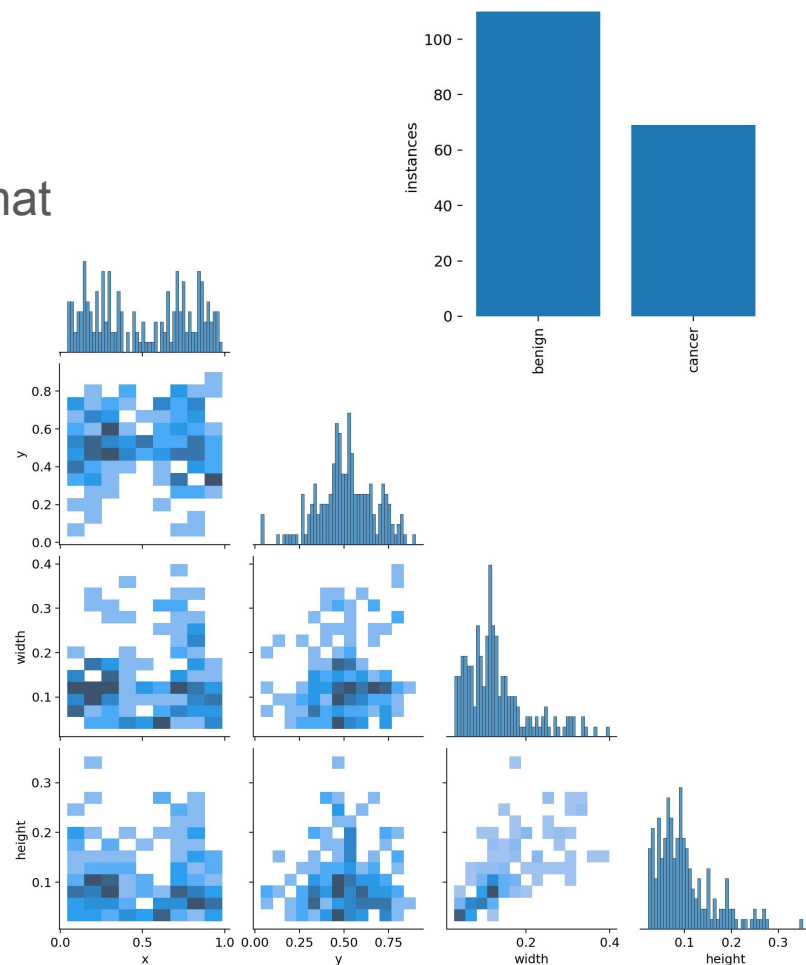
- Lags behind in accuracy
- Struggles with small objects

Method: Workflow



- Built the dataset and convert to YOLO format
- Two configuration files
 - Specify architecture to be used
 - Number of classes
- Train-test (0.9-0.1) split
 - $N_{\text{train}} = 179$, $N_{\text{test}} = 21$
- Model training
 - Labels: benign, cancer, (background)
- Model evaluation
 - Multiple metrics

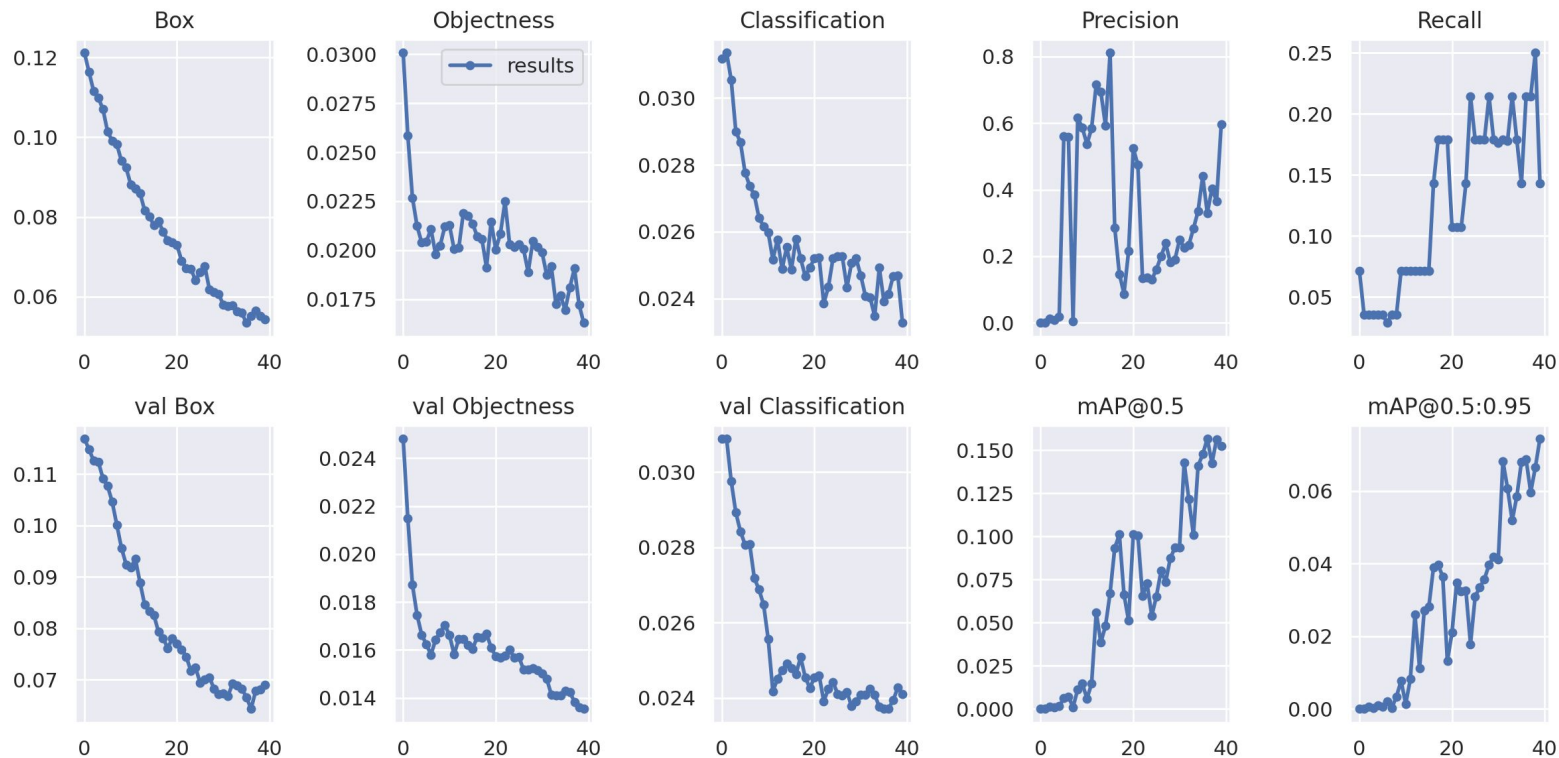
0 0.26064359295031164 0.12173736807287297 0.03206412825651302 0.026048026048026047



Preliminary results

Model: YOLOv5s (the smallest YOLOv5)

- Using their default architecture and parameters
- Image size 640
- Batch 16
- Epochs 40
- ~ 2hrs to train

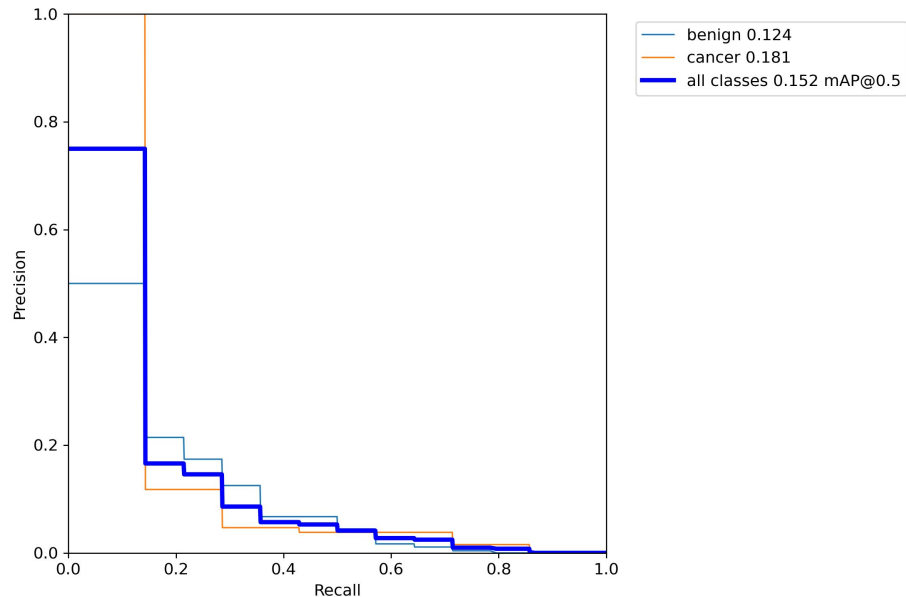


Preliminary results

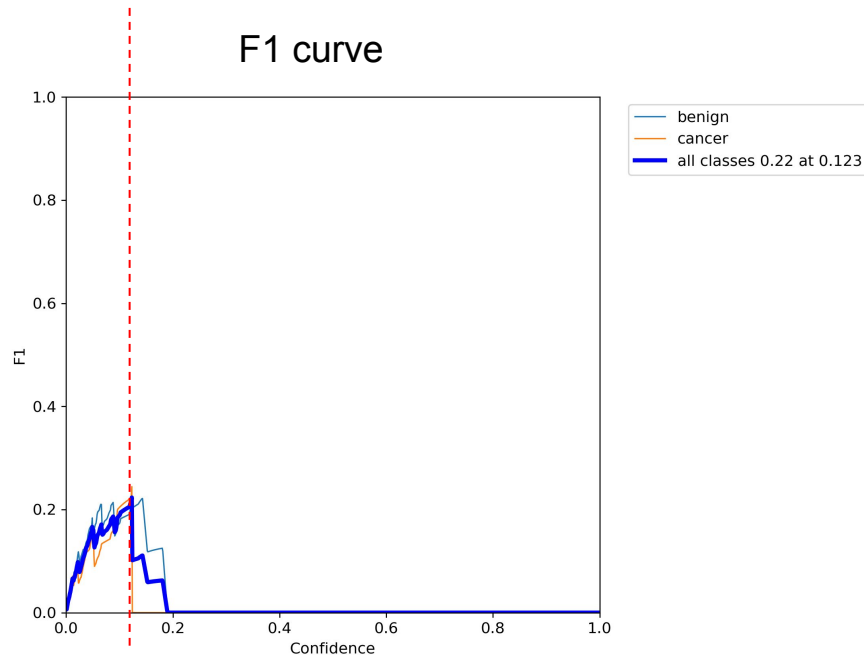
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Precision-recall curve



F1 curve

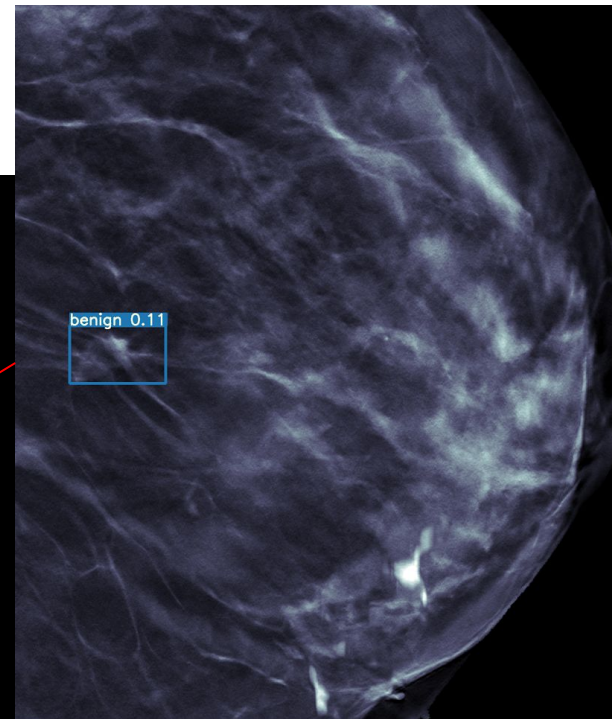
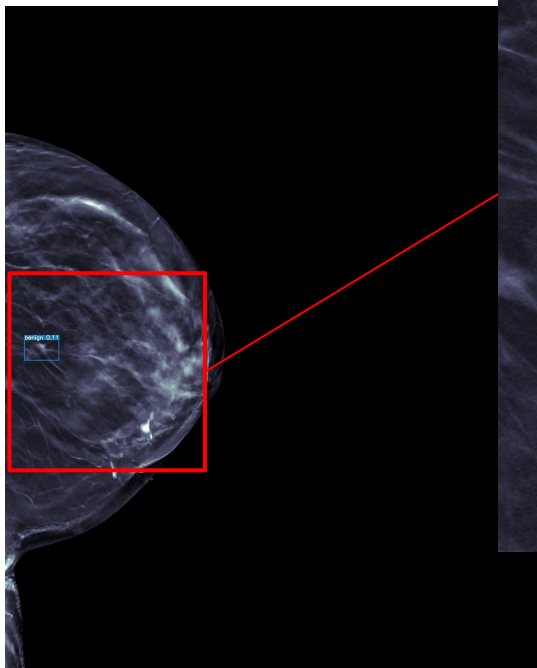
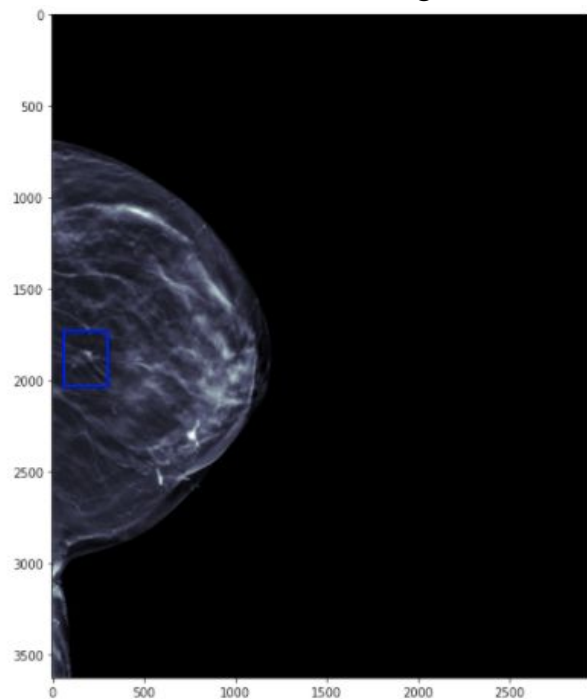


Preliminary results

Predict: YOLOv5s (the smallest YOLOv5)

- Using best weights trained
- Confidence 0.1

True label: benign

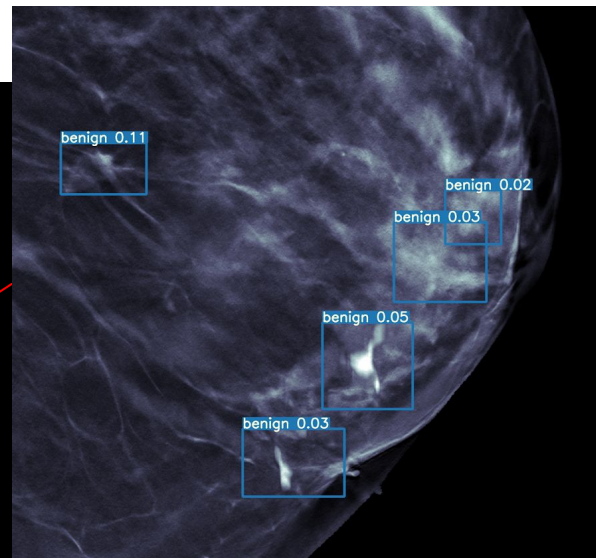
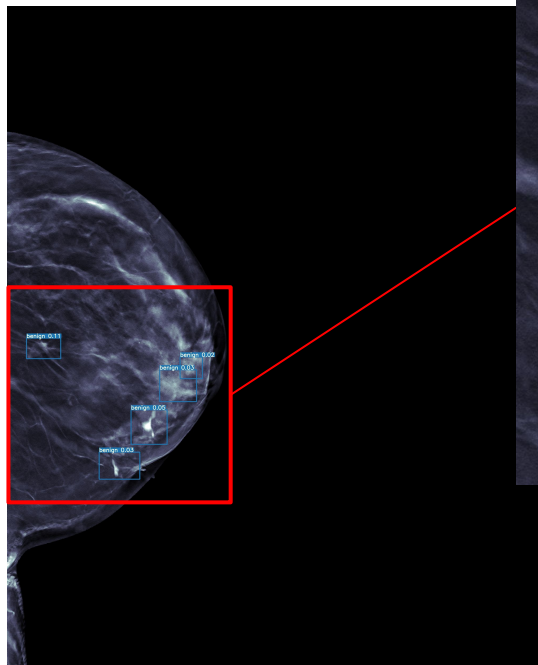
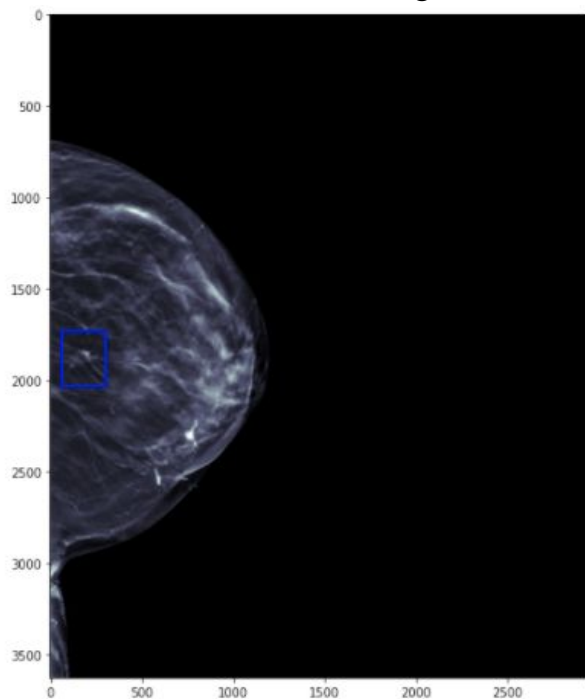


Preliminary results

Predict: YOLOv5s (the smallest YOLOv5)

- Using best weights trained
- Confidence 0.02

True label: benign

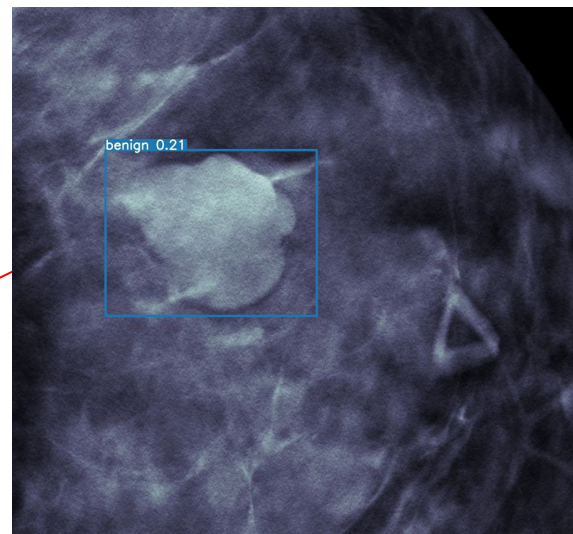
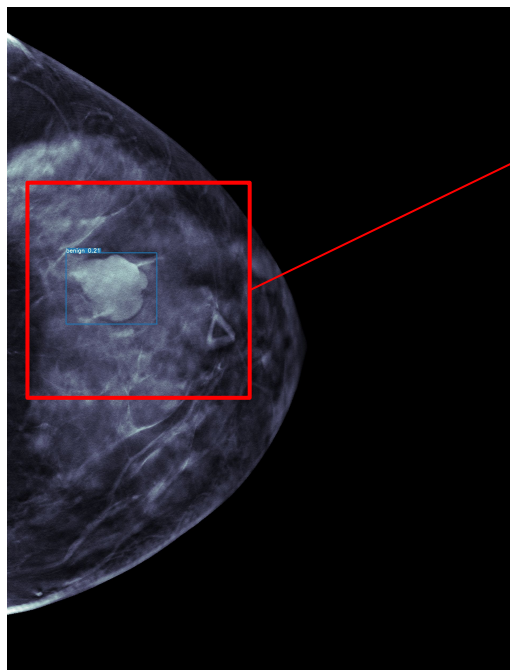
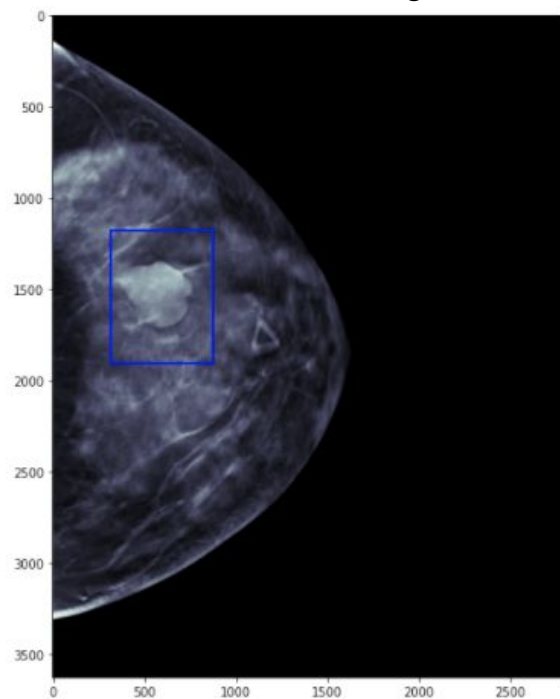


Preliminary results

Predict: YOLOv5s (the smallest YOLOv5)

- Using best weights trained
- Confidence 0.02

True label: benign

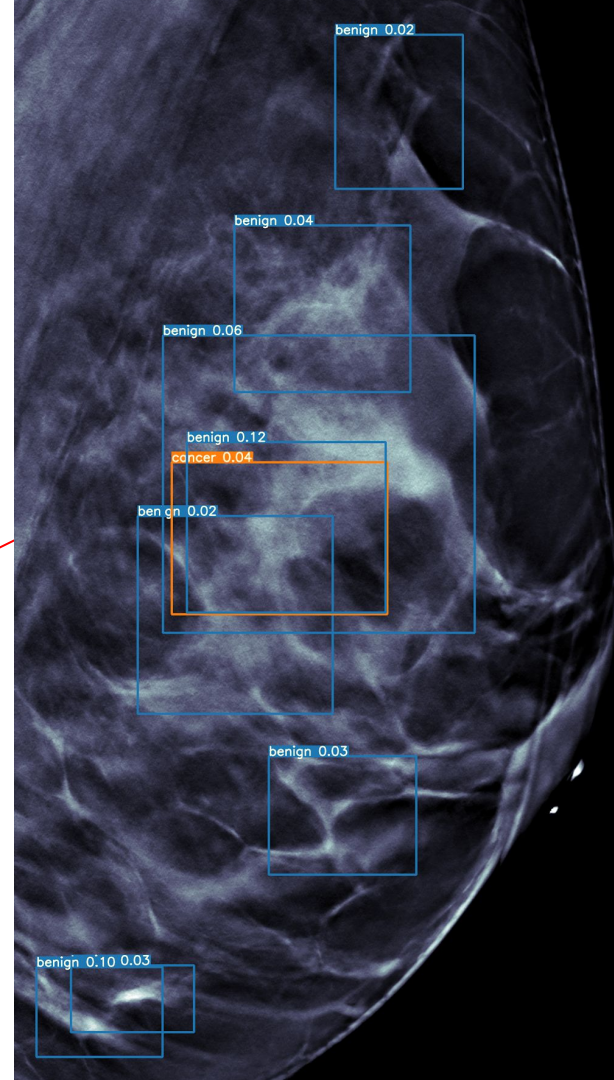
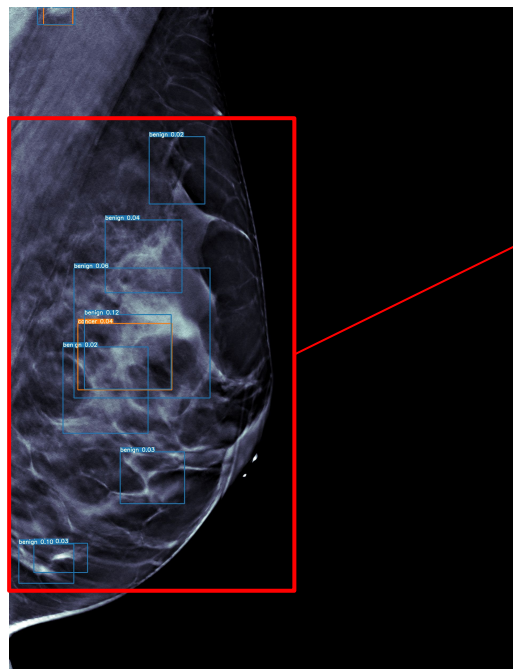
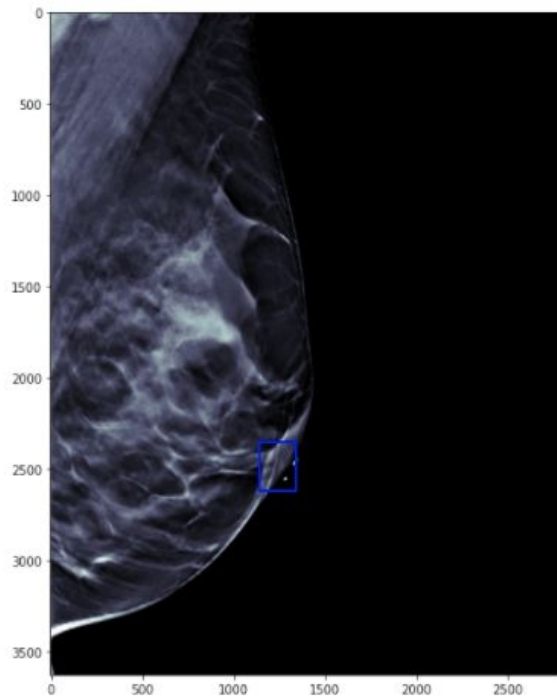


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Reflections and future works

- The current model is not ideal for ROI detection and classification
- The current train-test split doesn't include cancer in the test set
- Try larger YOLOv5 frameworks

