

Things done:

- 1) Getting the dataset ready for further use (like change of format)
- 2) Researched models that are feasible to use for our purpose
- 3) Testing of the models

Process in Detail:

- 1) Getting the dataset ready for further use (like change of format)

Upload the annotated data to account created in

<https://app.roboflow.com/>

Here we can do data augmentation, format conversion and also enable us to use it directly in platforms like python, c++ etc.

For the above purpose refer:

<https://www.youtube.com/watch?v=76E6esnez8E>

- 2) Researched models that are feasible to use for our purpose

Short listed some good models:

- (i) RCNN
- (ii) Mask RCNN
- (iii) Fast RCNN
- (iv) Faster RCNN
- (v) R-FCN
- (vi) YOLO
- (vii) Blitznet
- (viii) Retinanet

I decided not to use (i) as it is not very accurate, (ii) as we don't need masked region of the pre cancer

- 3) Testing the model

- (i) Faster RCNN:

There are a lot of different backbones to try here. I have tried the following :

fasterrcnn_darknet, fasterrcnn_mobilenetv3_large_fpn,
fasterrcnn_squeezezenet1_1, resnet50_fpn, resnet50_fpn_v2, vitdet,
vitdet_tiny, fasterrcnn_efficientnet_b0, fasterrcnn_efficientnet_b4

Out of the above only resnet101, resnet50_fpn and
fasterrcnn_efficientnet_b0 gave good results. Also see Vitdet.

(ii) Retinanet:

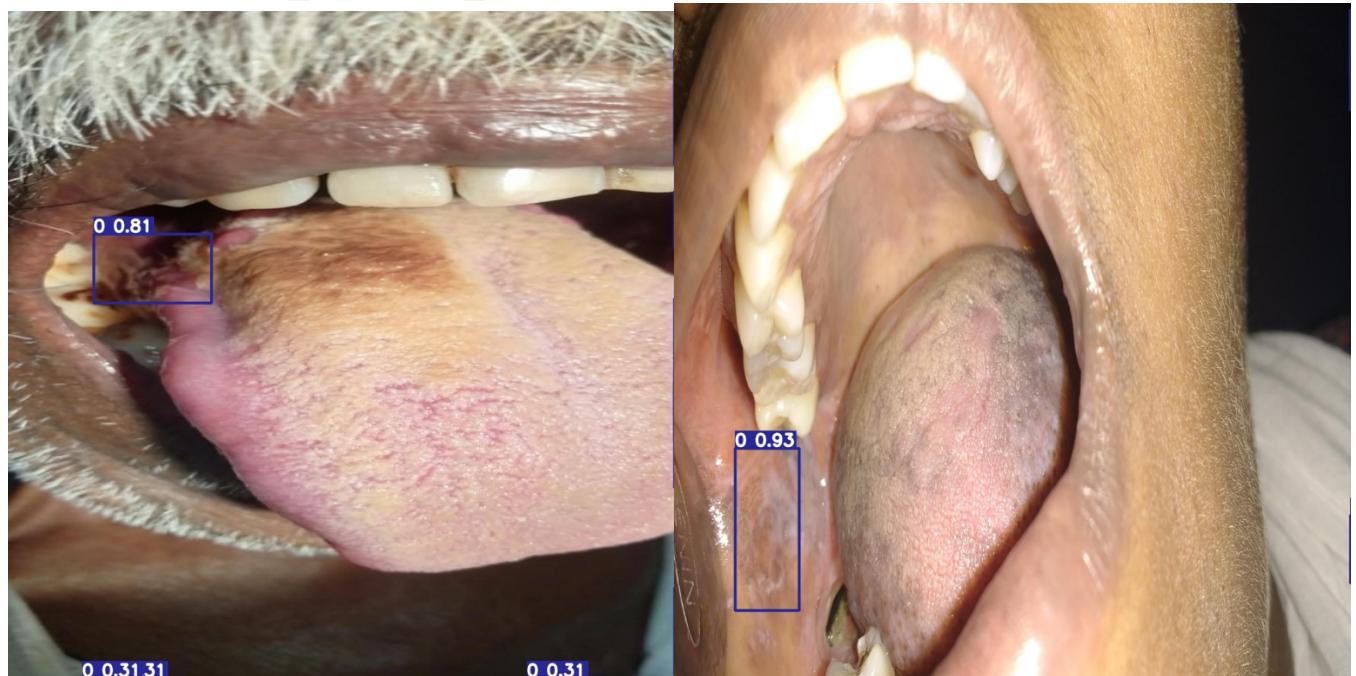
The result were not very good, but when precancer was detected it
was correctly.

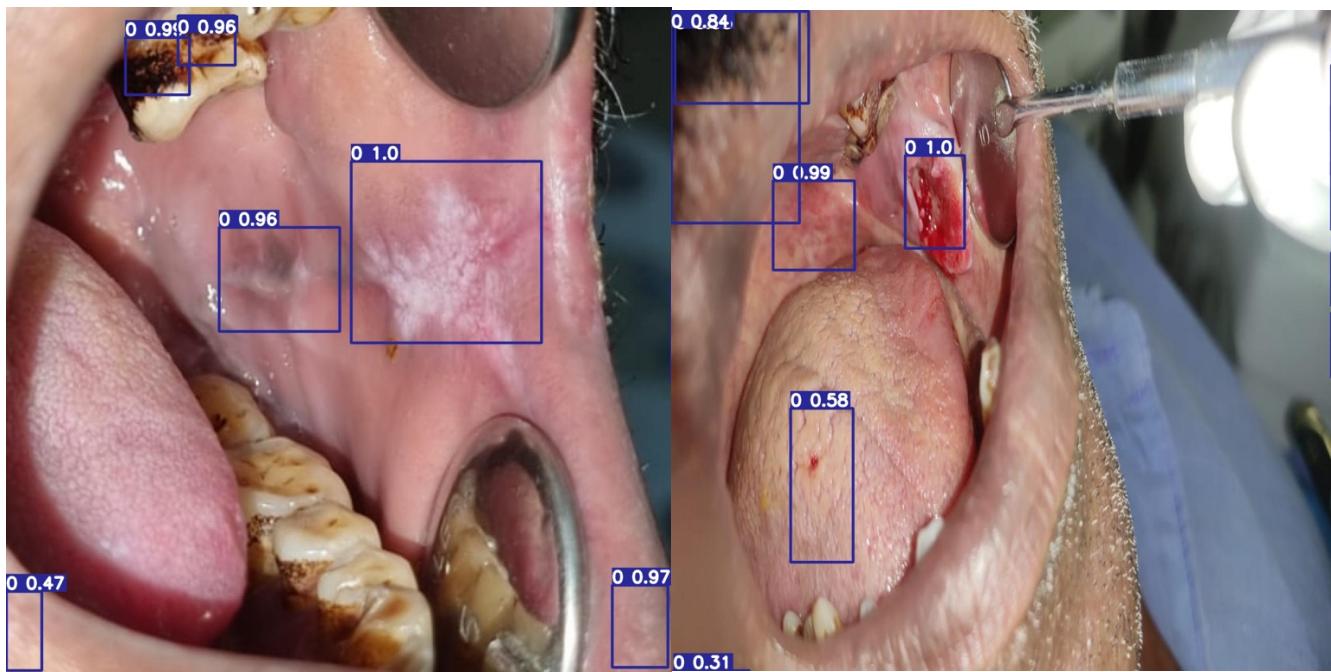
(iii) Yolov5:

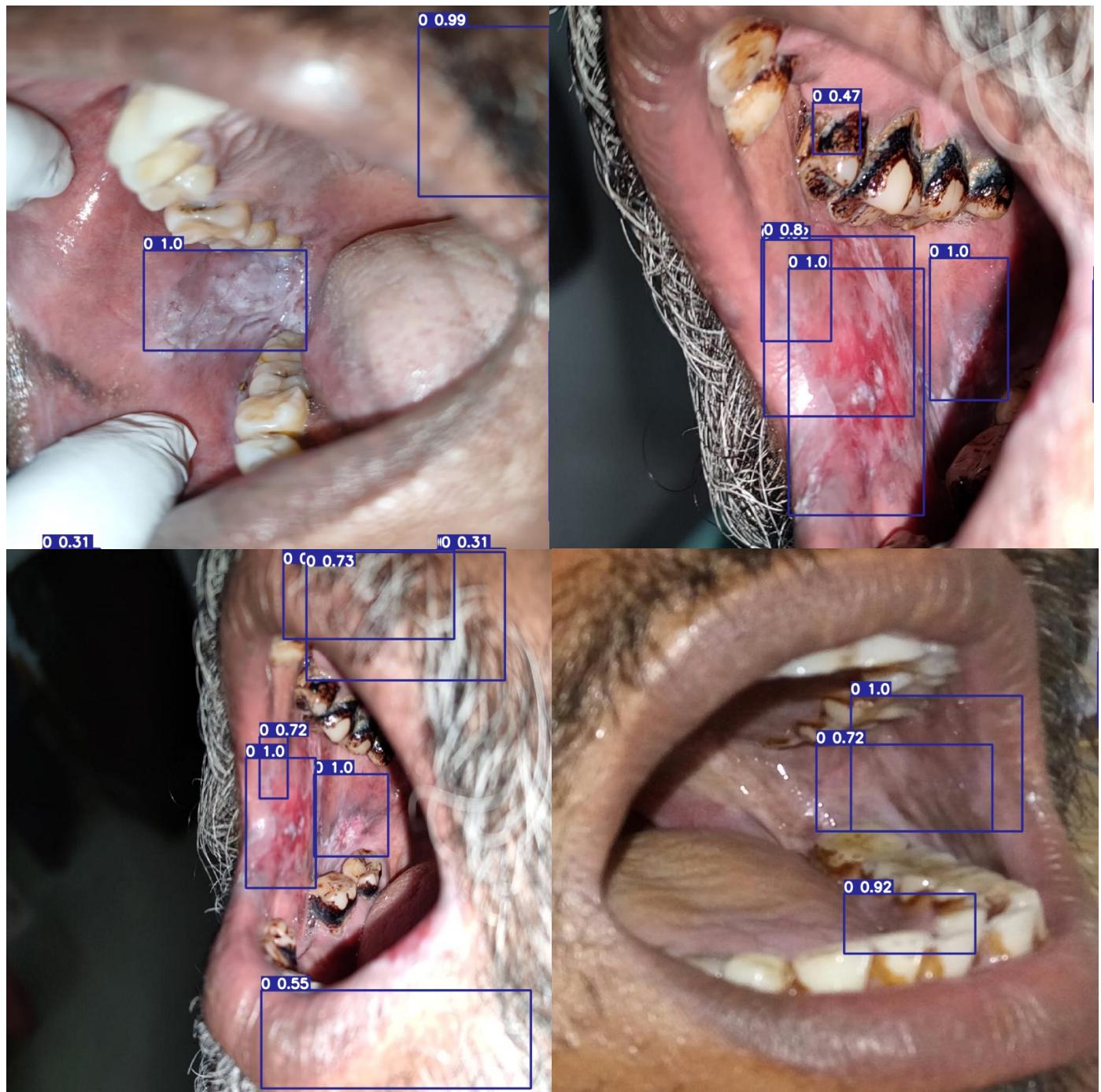
Accuracy was good but it also gave many false prediction.

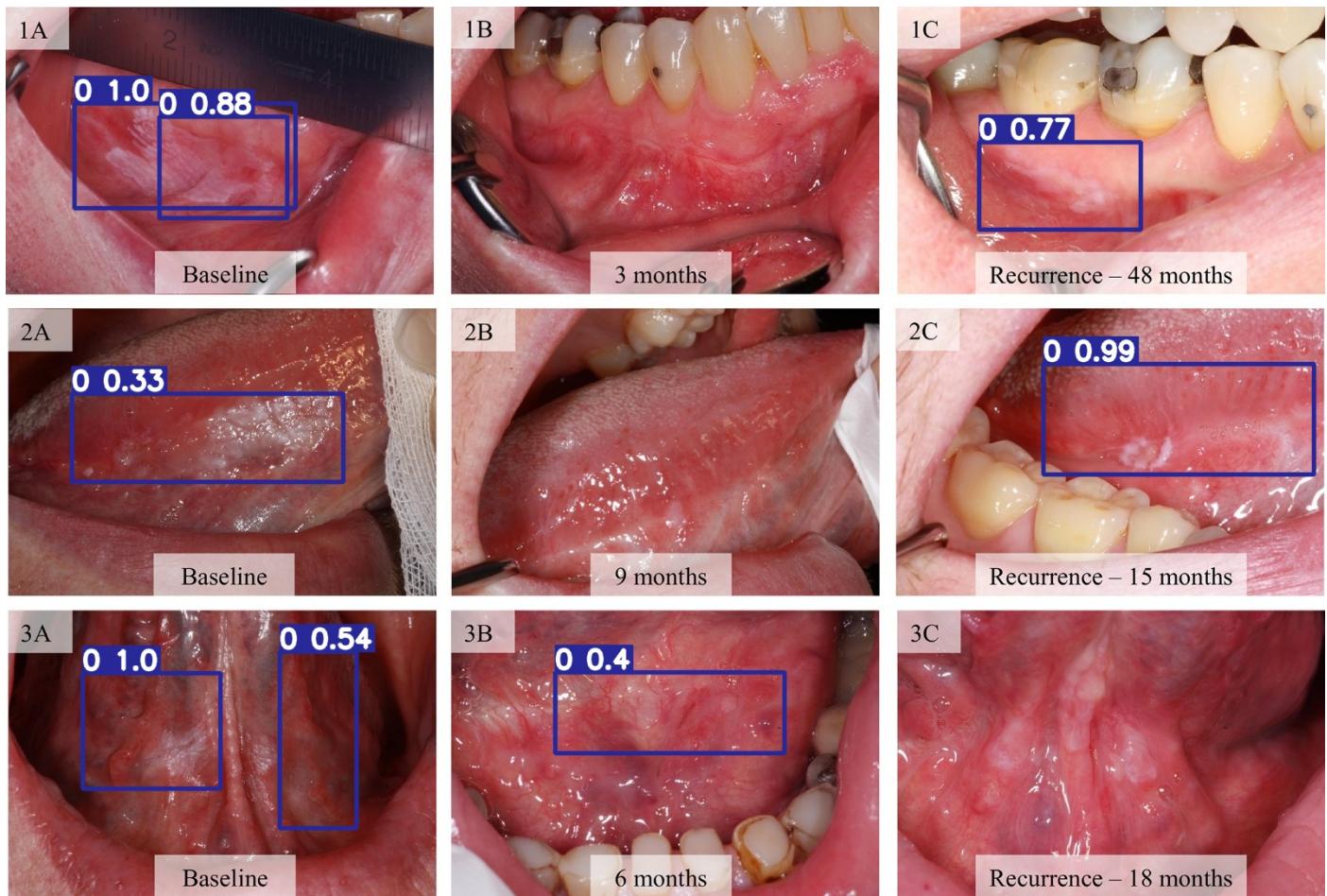
Results: (using 239 images in total)

Fasterrcnn_resnet50_fpn



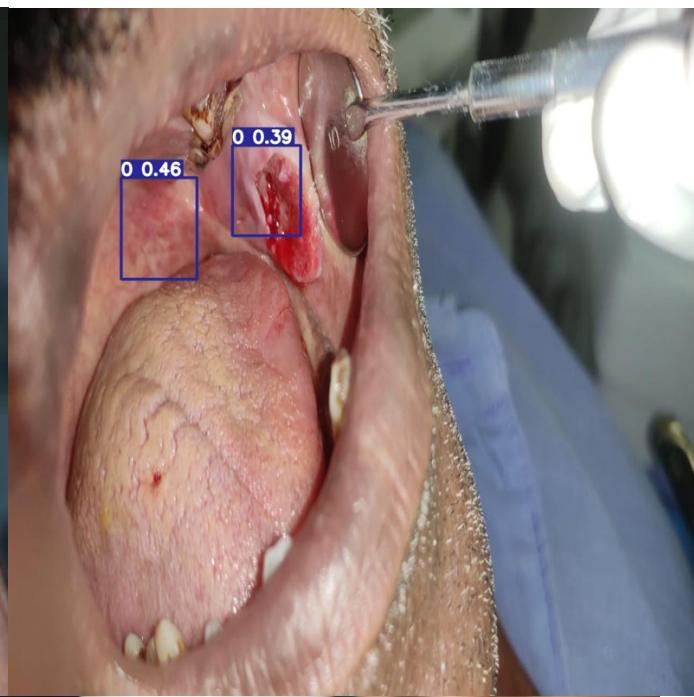


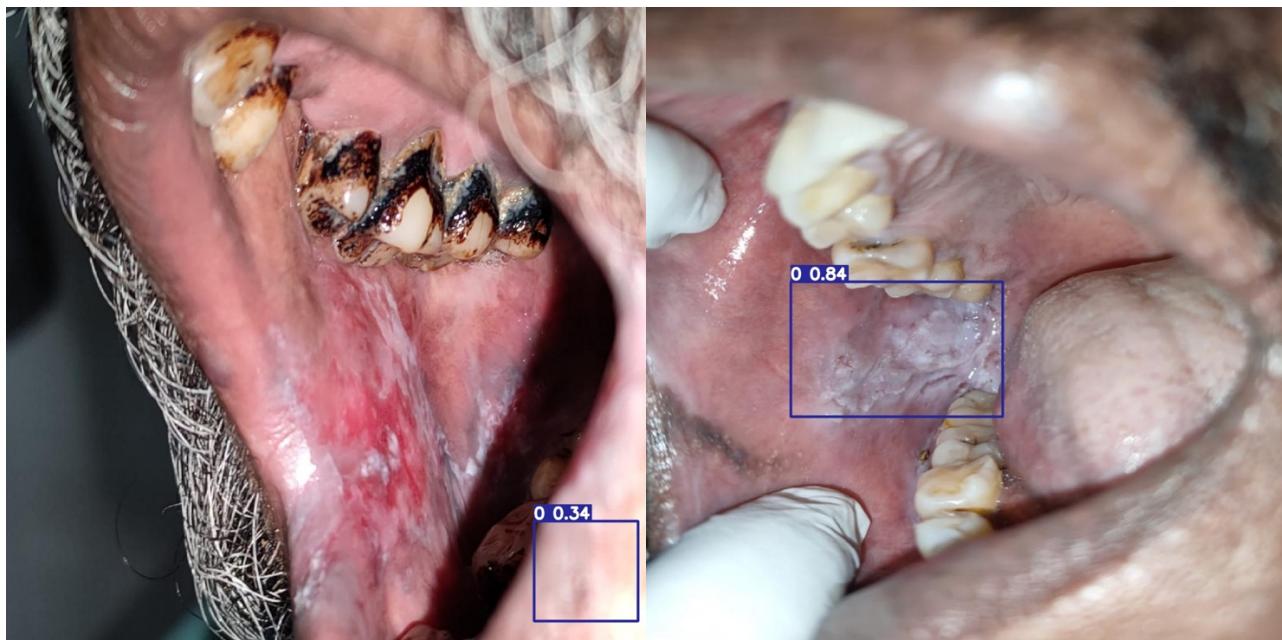




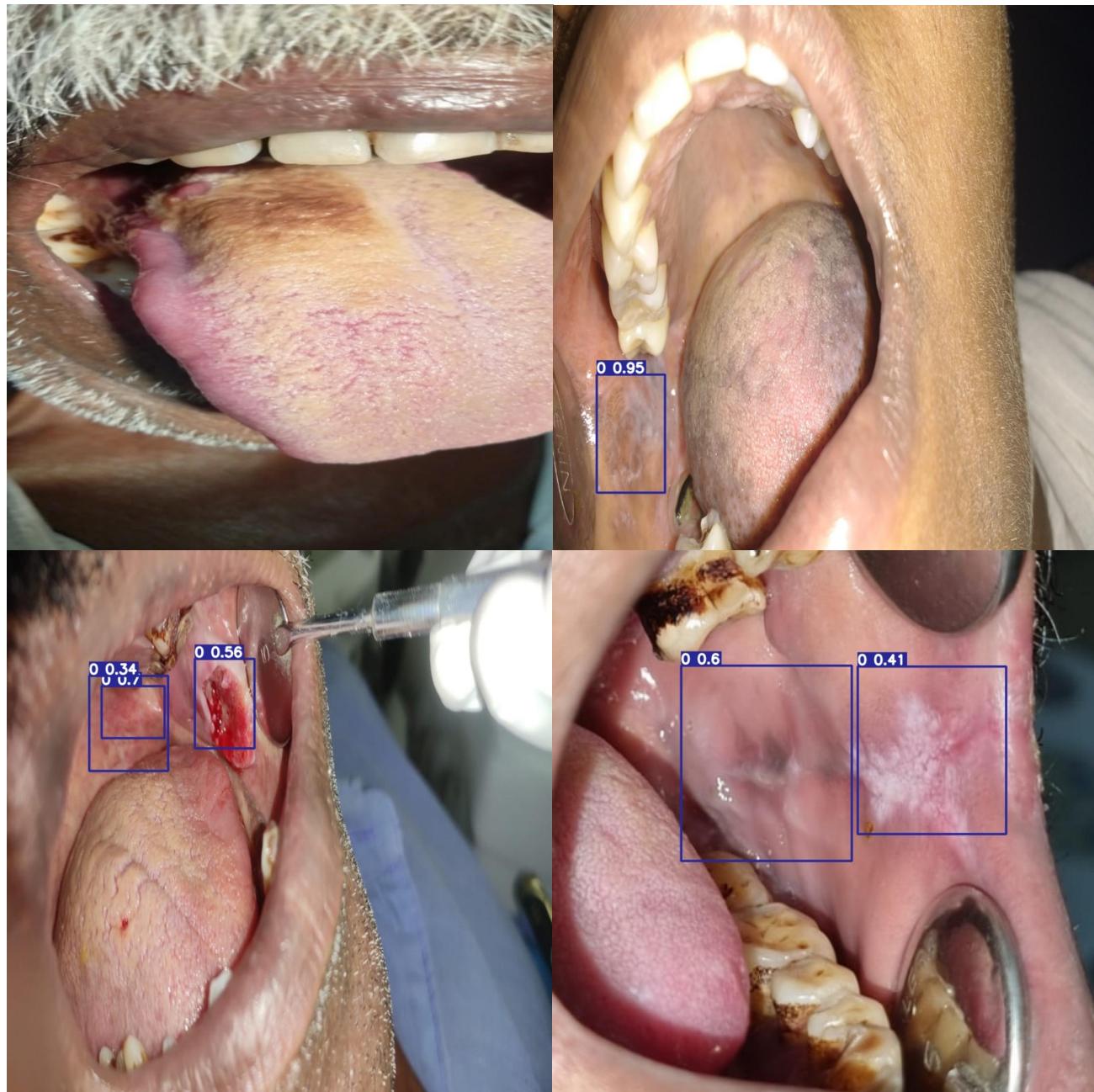
`fasterrcnn_resnet101`

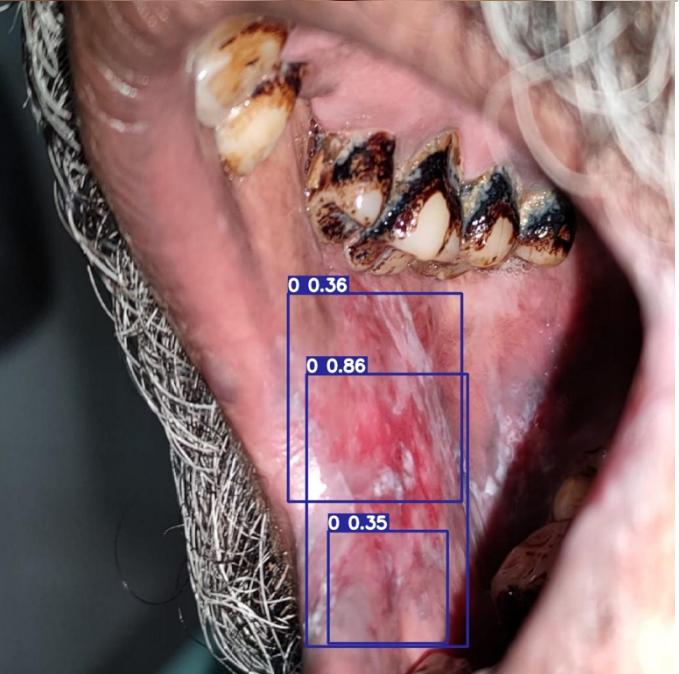
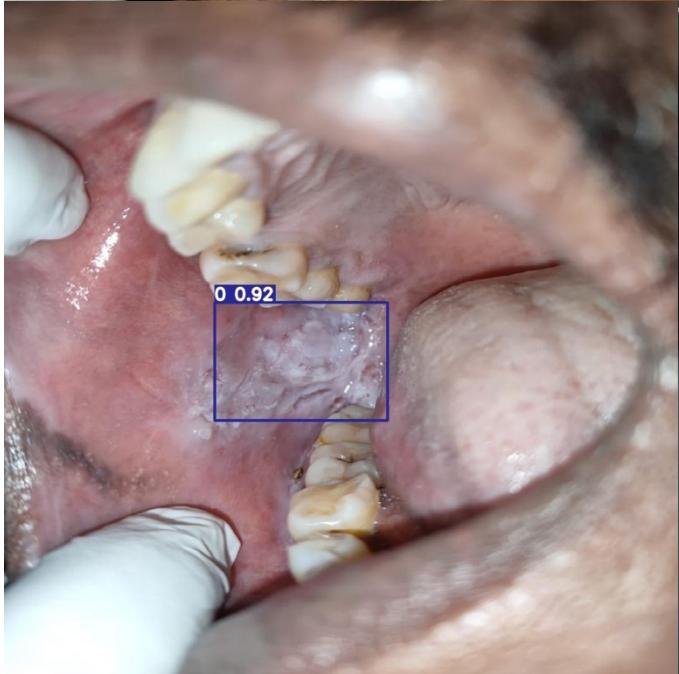


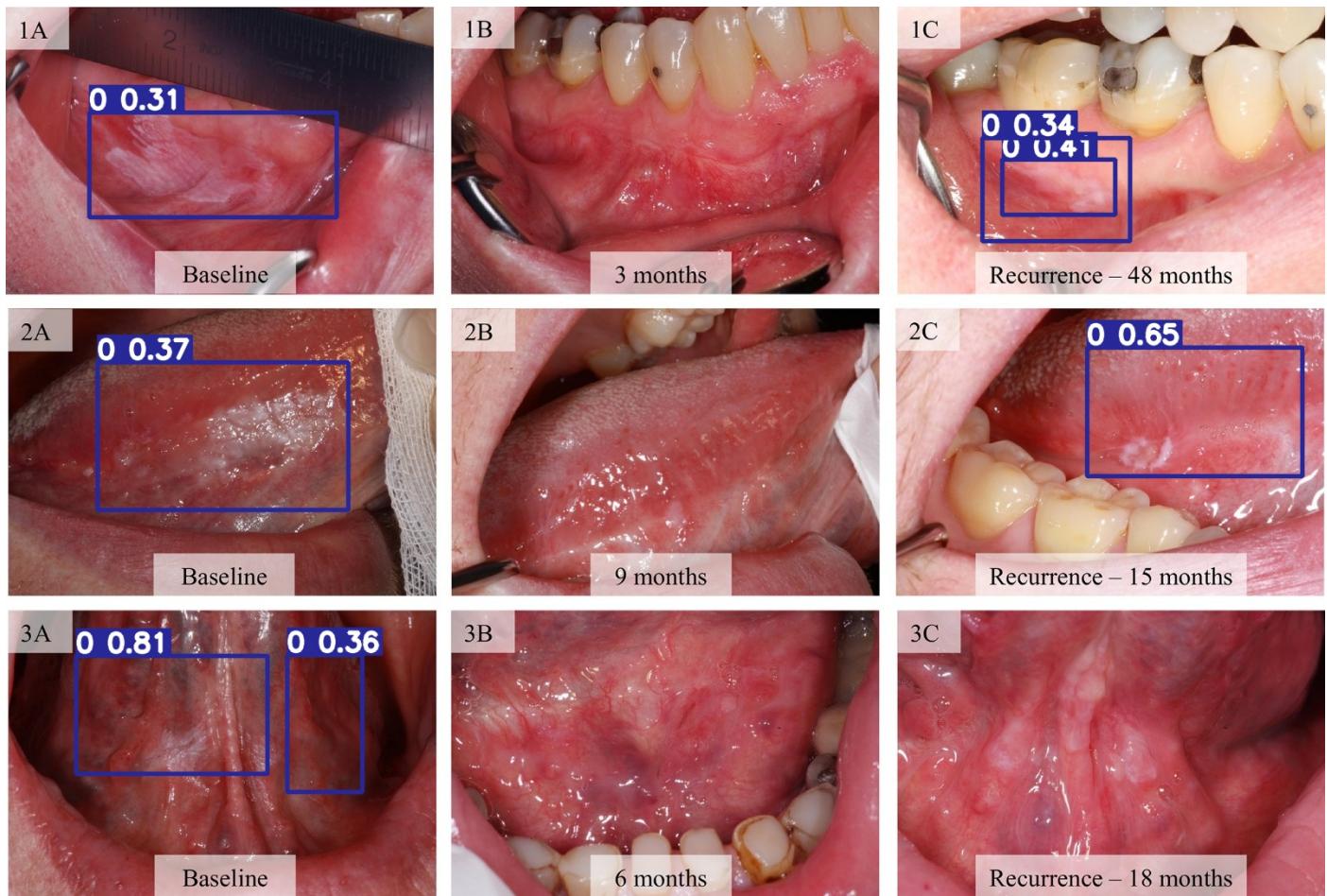




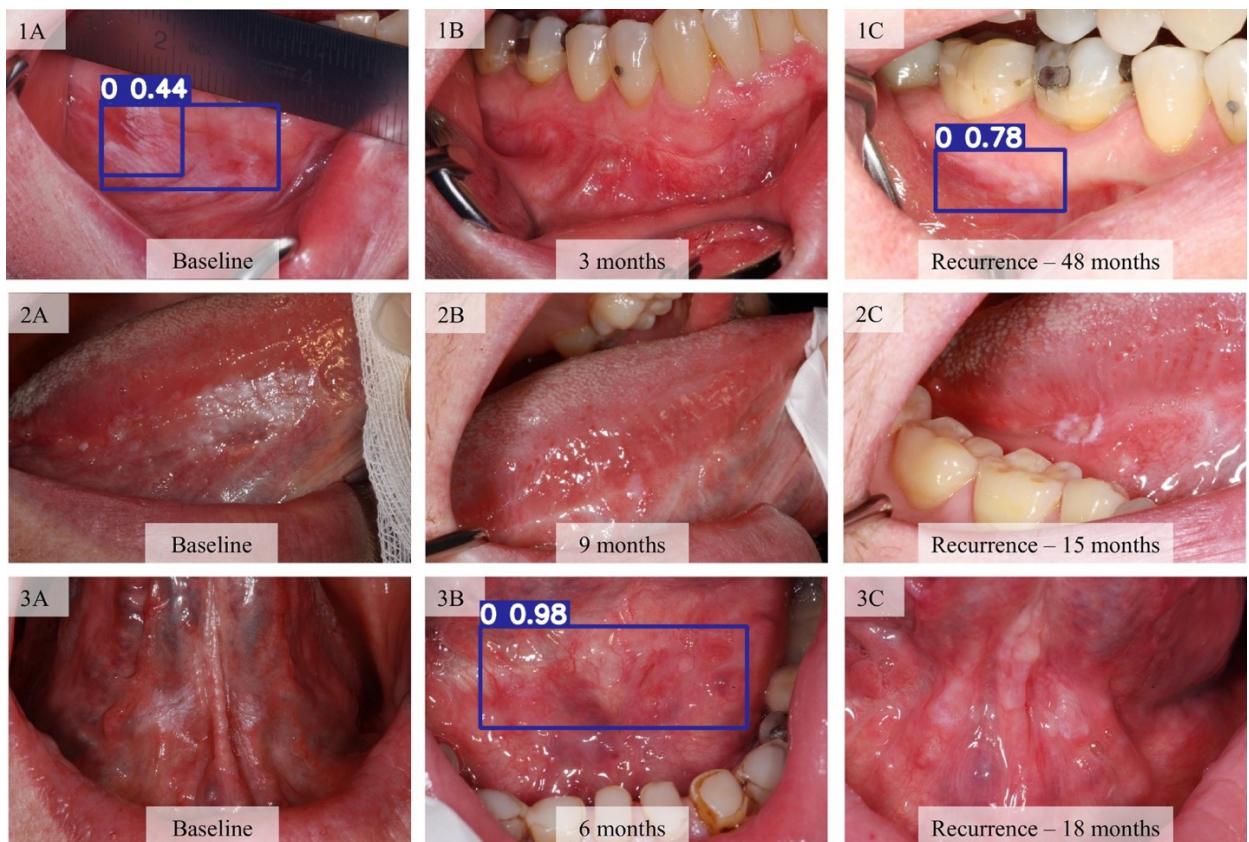
fasterrcnn_efficientnet_b0



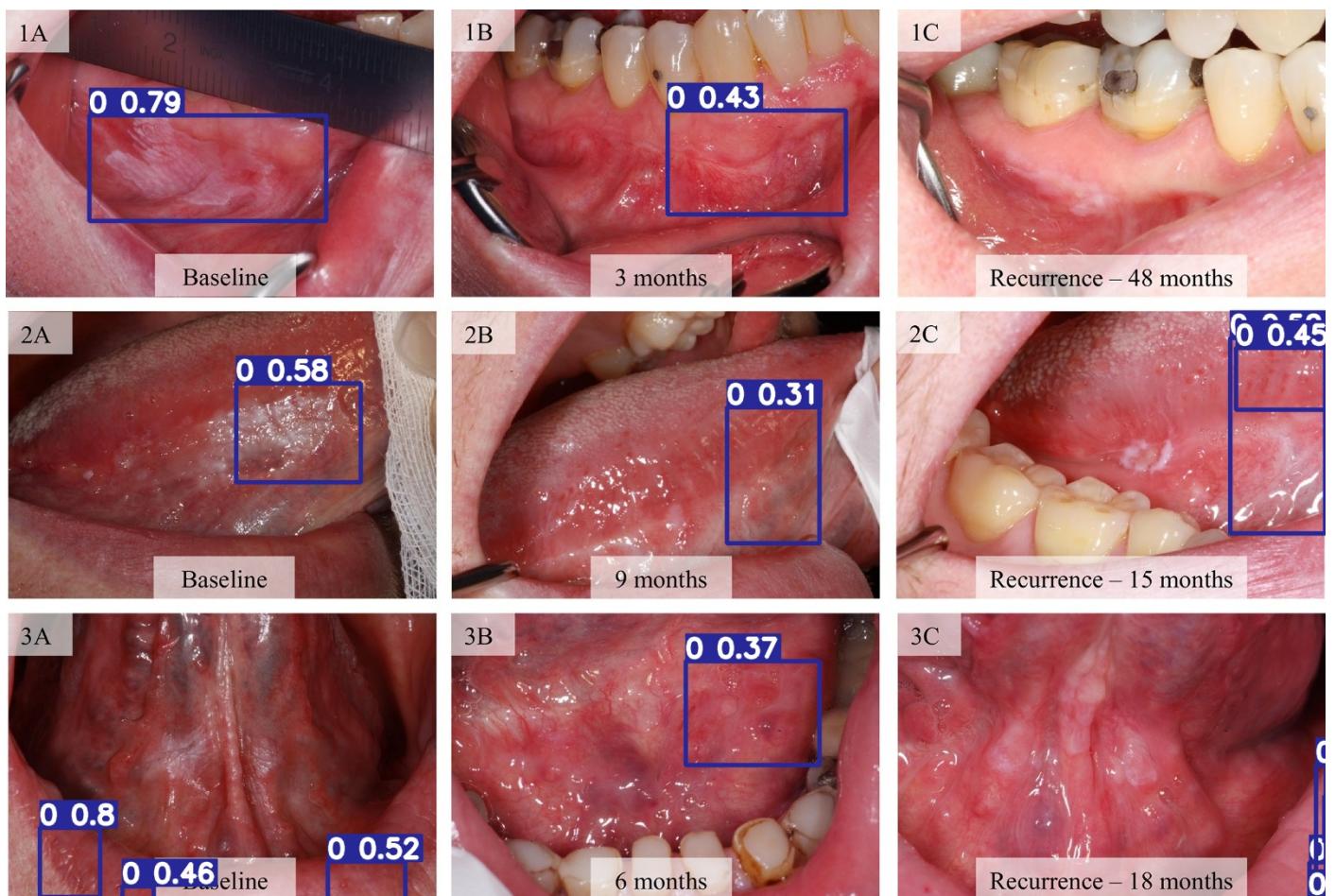




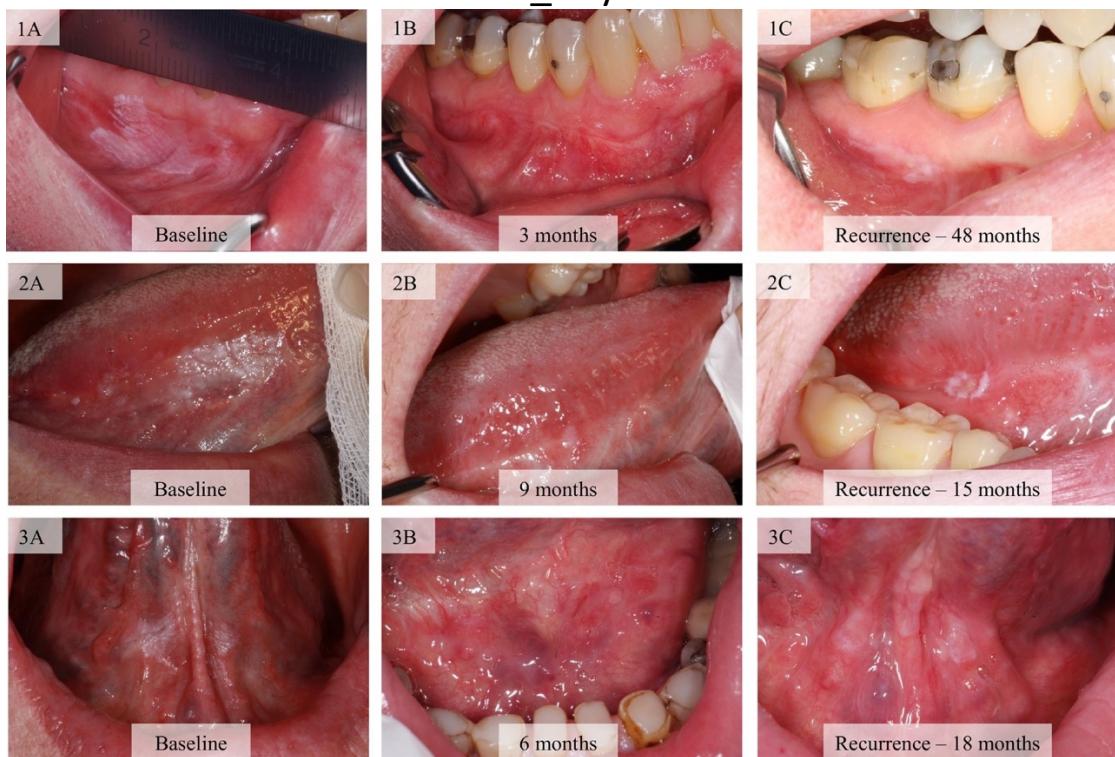
`fasterrcnn_mobilnetv3_large_fpn`



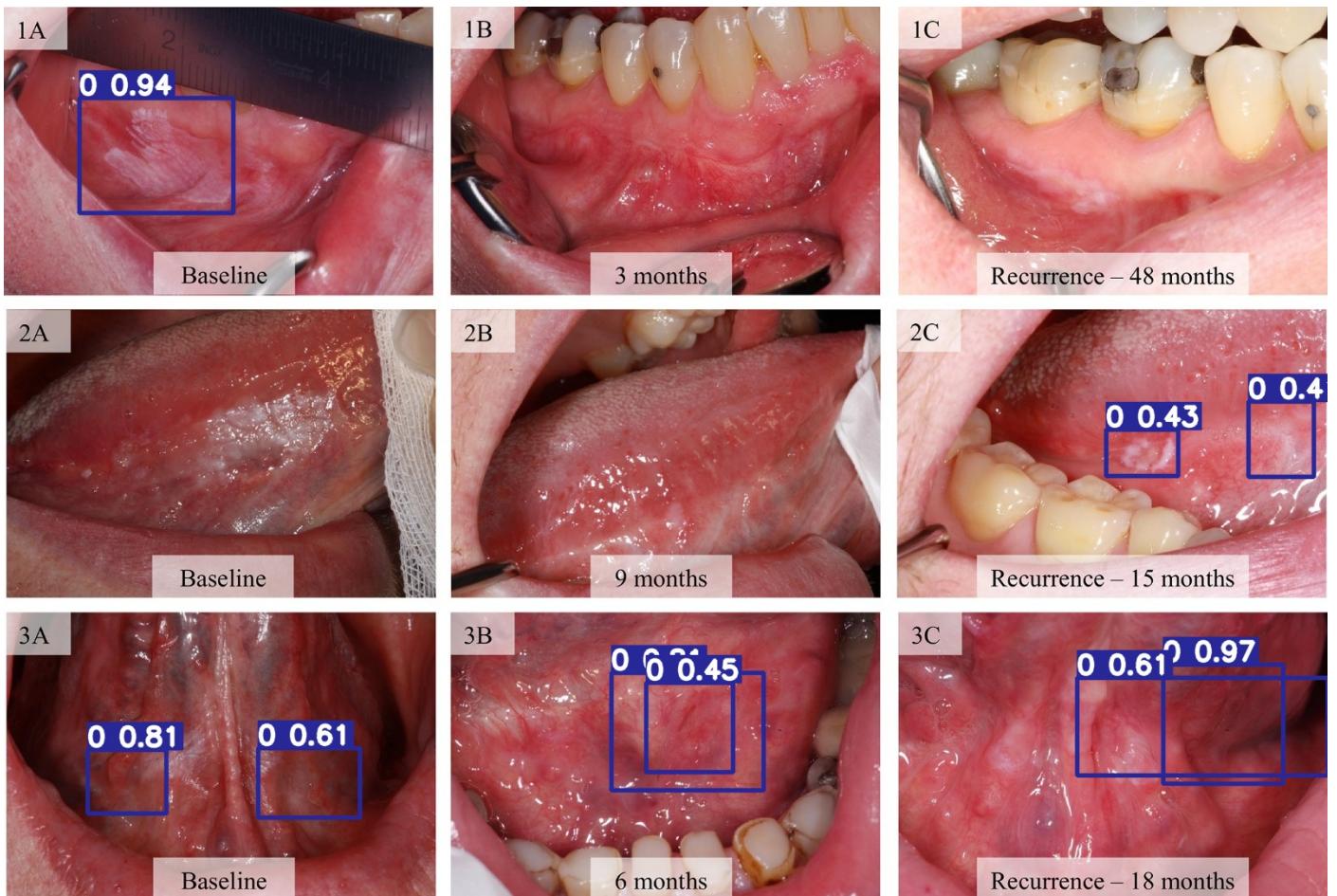
fasterrcnn_squeezeNet1_1



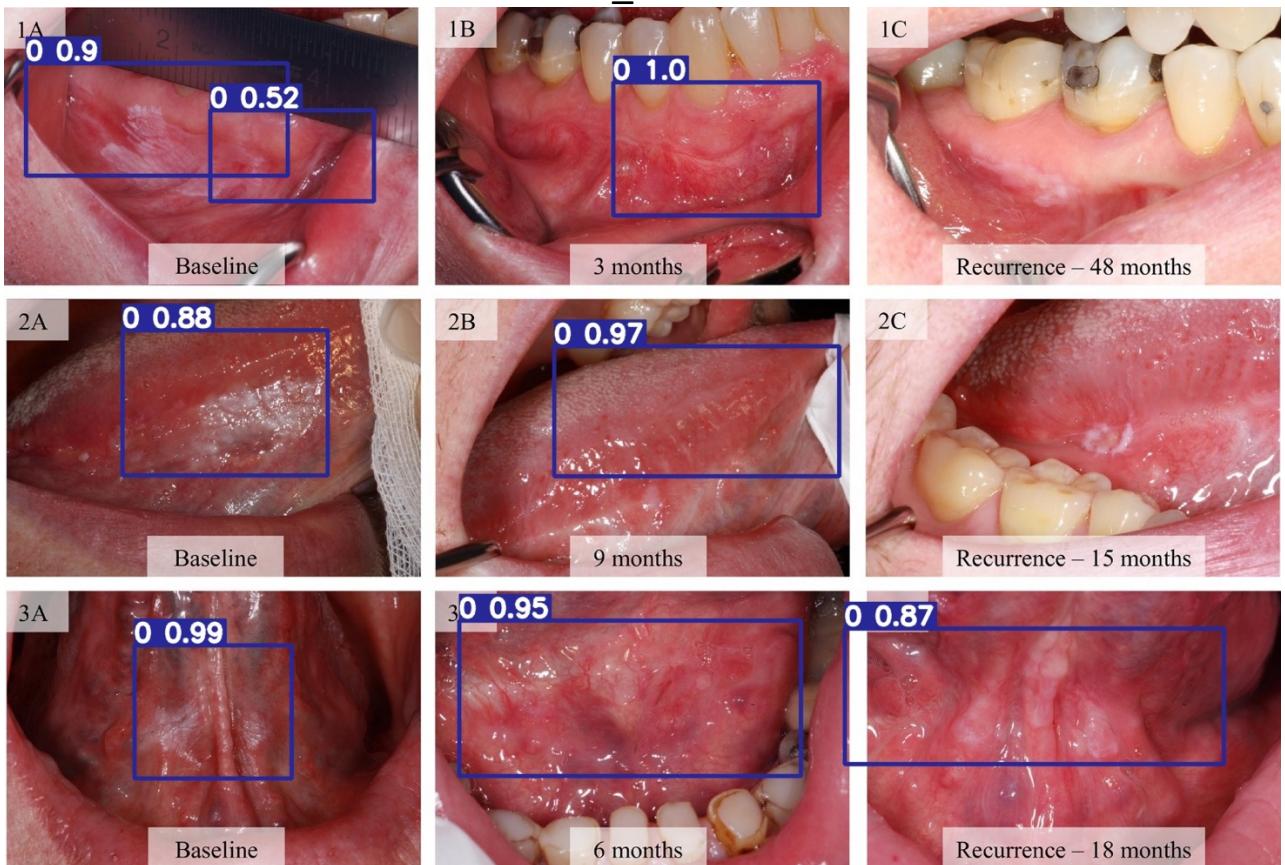
vitdet_tiny



Fasterrcnn_vitdet



Fasterrcnn_darknet



Procedure for various models:

i) **Retina net**

To train :

```
!python train.py --dataset coco --coco_path <path of .coco file of the dataset> --depth <depth of network>  
--epochs <number of epochs>
```

To test a coco dataset:

```
!python visualize.py --dataset coco --coco_path <path of .coco file of the dataset> --model <path of .pt file  
of weights obtained during training>
```

To test a file of images:

```
python visualize_single_image.py --image_dir <directory of test images> --model_path <.pt file path>  
--class_list <.csv of classlist path>
```

what is classlist : we have to make a csv with classes of the following format,
class_key, class_name . Also include 0,background.

Visit : <https://github.com/yhenon/pytorch-retinanet> for further information

The data should of a .json format and the names should be very specific:

Steps:

- 1) Create file named “oral_precancer.v1i.coco” inside the cloned repo.
- 2) Make two directory “annotations”,“images”
- 3) In “annotations” : rename annotation of train data as “instances_train2017.json” and of validation as “instances_val2017.json”
- 4) In “images” : store train images in “train2017” and validation images in “val2017”

ii) **Faster RCNN**

To train:

```
python train.py --data <location of .yaml file of the json format data> --epochs <number of epochs>  
--model <model name> --name <name of file in which to save the results> --batch <batch size>
```

To test:

```
python inference.py --input <location of directory of test images> --weights <path of .pth either  
best_model.pth or last_model_state.pth>
```

the results of training are saved at ./outputs/training file

the inferences are saved at ./outputs/inference

to find all available models check ./models

go to /home/aimlgpu2/yash/oral_pre_cancer/fastercnn-pytorch-training-pipeline on the server to get an idea as it is already implemented there.

For further info on this visit : <https://github.com/sovit-123/fastercnn-pytorch-training-pipeline>
It also contains info on how to make the .yaml file.

You can also change the model used in efficient net file by going to fastercnn_efficientnet_b0.py but you will also have to change the “backbone.out_channels =” (for b0,b1 : 1280, b2:1408, b6:2304 , you can find this value from error that is thrown if you can model and don’t change this value).

You can also freeze the weights of backbone:

```
for param in model.backbone.parameters():
    param.requires_grad = False
```

Just add the above line

You can find all of this stuff on this folder on gpu2 server : “/home/aimlgpu2/yash/oral_pre_cancer”, ip : sftp://192.168.15.5