ITI107 Assignment Object Detection

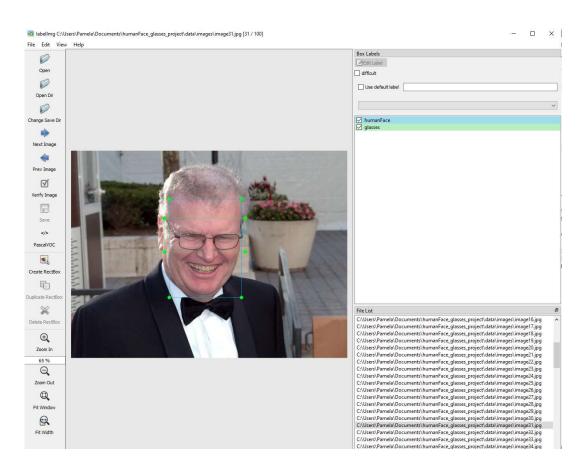
Pamela Sin Hui (21A530M)

Section 1: Data Collection and Annotation

A pip install of the fiftyone library was done locally. The below code was run, where 150 images containing objects of "Glasses" and "Human face" was downloaded from the Open Images dataset (https://storage.googleapis.com/openimages/web/download.html).

From the 150 images, the first 100 were taken to be our final dataset, where subsequently specific images were swapped out, such as in an instance where an image showcased a large crowd of people which would have made annotation difficult. The 100 images were renamed from image1.jpg to image100.jpg for consistency. As "Glasses" and "Human face" objects tend to go hand-in-hand and be in the same image (i.e. the image contains a person wearing glasses), each of the 100 images have at least one "Glasses" object and one "Human face" object in the image.

Each of the 100 images was manually annotated using the LabelImg tool, using the PascalVOC format. For the annotation of "Glasses" objects, both clear glasses and sunglasses were annotated, and the entire object including the temples are being included in the ground truth label. For the "Human face" objects, only the human face, excluding the hair and ears are being included in the ground truth label. The images are taken from a variety of settings, be it indoors or outdoors, as it is the intention for the object detection model to be able to detect objects in different settings.



Section 2: Training Process

Training was done following step-by-step instructions from https://github.com/nyp-sit/iti107/blob/main/session-4/custom_training_with_tfod_api.md

Firstly, the images and annotation files are copied to separate folders. Next, the label map (.pbtxt file) of each numeric label to its corresponding text label was created. Subsequently, the images and annotations data was converted to TF Records binary format, with a train to validation split of 80% to 20%.

The pretrained model selected was ssd_resnet101_v1_fpn_640x640_coco17_tpu-8, given its relatively higher COCO mAP versus speed in comparison with other SSD models

(https://github.com/tensorflow/models/blob/master/research/object_detection/g3doc/tf2_detection_zoo.md).

SSD MobileNet v2 320x320 SSD MobileNet V1 FPN 640x640 SSD MobileNet V2 FPNLite 320x320 SSD MobileNet V2 FPNLite 640x640 SSD ResNet50 V1 FPN 640x640 (RetinaNet50) SSD ResNet50 V1 FPN 1024x1024 (RetinaNet50)	19 48 22 39	20.2 29.1 22.2 28.2	Boxes Boxes Boxes
SSD MobileNet V2 FPNLite 320x320 SSD MobileNet V2 FPNLite 640x640 SSD ResNet50 V1 FPN 640x640 (RetinaNet50)	22	22.2	Boxes
SSD MobileNet V2 FPNLite 640x640 SSD ResNet50 V1 FPN 640x640 (RetinaNet50)	22	PERMED	1270 Mari
SSD ResNet50 V1 FPN 640x640 (RetinaNet50)	39	28.2	2000
esterni de considera do como o como como formación de la como de como de la como de co			Boxes
SSD ResNet50 V1 FPN 1024x1024 (RetinaNet50)	46	34.3	Boxes
	87	38.3	Boxes
SSD ResNet101 V1 FPN 640x640 (RetinaNet101)	57	35.6	Boxes
SSD ResNet101 V1 FPN 1024x1024 (RetinaNet101)	104	39.5	Boxes
SSD ResNet152 V1 FPN 640x640 (RetinaNet152)	80	35.4	Boxes
SSD ResNet152 V1 FPN 1024x1024 (RetinaNet152)	111	39.6	Boxes

To perform hyperparameter tuning, the object detection pipeline file was configured differently for separate runs of the training. A batch size of 4 was used for all runs due to limited GPU memory.

Section 3: Experimental Results

Run 2 contains the standard default hyperparameters defined in the pretrained model's configuration file.

For Run 3, the first anchor aspect ratio tweaked was from 2.0 to 1.5. This newly selected arbitrary aspect ratio seems to better capture the typical height to width ratio of the object "Human face", making this anchor less elongated. The second anchor aspect ratio tweaked was from 0.5 to 0.4, where this aspect ratio appears to better encompass the object "Glasses" sitting on a person's face is slightly wider. In comparing the mAP@IoU=0.50 between Run 2 and Run 3, Run 2 performed better at 0.643 versus Run 3 at 0.604.

In considering the experimental results of Run 2 and Run 3, for Run 4, Run 2's anchor aspect ratios was used, while the classification weight to localization weight was tweaked from 1.0:1.0 to 1.2:1.0. This was done as Run 2's Tensorboard plots of classification loss against number of batches hovers around 0.10 towards the tail end of training, while that for the localization loss is around 0.05. Hence in Run 4, we want to determine if placing greater weights on classification versus localization in the loss function will improve the performance metric. However, in comparing the mAP@IoU=0.50 between Run 2 and Run 4, Run 2 performed better at 0.643 versus Run 3 at 0.628.

We note that training for an even greater number of batches of data across the Runs could have led to different performance metric results. However, due to constraints in allocated GPU hours, we select the best model based on the experimental Runs thus far.

Section 3.1: Summary of Results

Run #	Run 2	Run 3	Run 4
Classification			
weight:			
Localization			
weight	1.0:1.0	1.0:1.0	1.2:1.0
Anchor Aspect			
Ratios	1.0, 2.0, 0.5	1.0, 1.5, 0.4	1.0, 2.0, 0.5
Step eval			
metrics are			
performed	11700	12900	12700
mAP@IoU=0.			
50	0.643	0.604	0.628
	Del 06 al 28 de noviembre de noviembre	Del 06 al 28 de noviembre Del 06 al de noviem	Del 06 al 28 de noviembre de noviembre
	WWW.LosMocanos.com	www.losMocanos.com	

Section 3.2: Run 2 Detailed Results

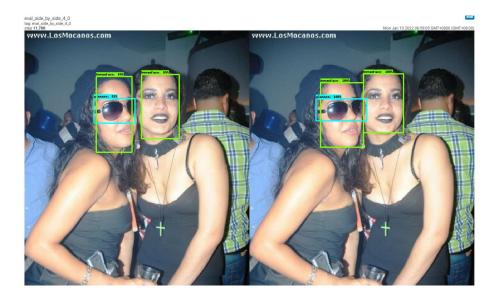
The graphs of classification loss, localization loss, regularization loss and total loss over number of batches all have a general downward trend, where as the number of training batches increases, the losses decreases at a decreasing rate and plateaus at approximately 9000 training batches. This general trend is also seen across runs 3 and 4.

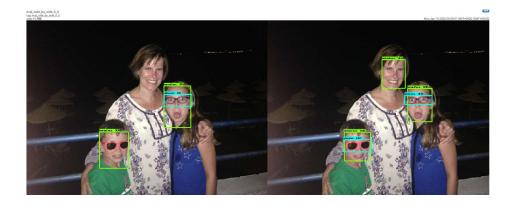
As Run 2 has the highest mAP@IoU=0.50, the side-by-side comparison of a few test images against the ground truth label is additional provided in this section.

```
(AP) @[ IoU=0.50:0.95
                                            area=
                                                           maxDets=100 1 = 0.314
Average Precision
                    (AP)
                            IoU=0.50
 Average Precision
                                                    all
                                                           maxDets=100
                                            area=
                                                                         = 0.643
                         @[
 Average Precision
                    (AP)
                         @[
                            IoU=0.75
                                                    all
                                                           maxDets=100
                                            area=
                                                                         = 0.183
 Average Precision
                    (AP)
                            IoU=0.50:0.95
                                            area= small
                                                           maxDets=100 ]
                                                                         = 0.000
                         @ [
                    (AP)
Average Precision
                            IoU=0.50:0.95
                                            area=medium
                                                           maxDets=100
                                                                         = 0.200
                         @[
 Average Precision
                    (AP)
                        @ [
                            IoU=0.50:0.95
                                            area= large
                                                           maxDets=100 ] = 0.342
 Average Recall
                    (AR) @[
                            IoU=0.50:0.95
                                            area=
                                                   all
                                                           maxDets= 1
                                                                       1 = 0.200
 Average Recall
                    (AR)
                            IoU=0.50:0.95
                                            area=
                                                    all
                                                           maxDets= 10
                                                                         = 0.396
                        @[
 Average Recall
                    (AR) @[ IoU=0.50:0.95
                                            area=
                                                    all
                                                           maxDets=100
                                                                       1 = 0.438
                    (AR)
 Average Recall
                         @[
                            IoU=0.50:0.95
                                            area= small
                                                           maxDets=100
                                                                       ] = 0.000
 Average Recall
                    (AR) @[ IoU=0.50:0.95
                                            area=medium
                                                           maxDets=100 ] = 0.331
 Average Recall
                    (AR) @[ IoU=0.50:0.95
                                                           maxDets=100 ] = 0.463
                                            area= large
INFO:tensorflow:Eval metrics at step 11700
I0109 22:59:20.827951 140110343554496 model_lib_v2.py:1007] Eval metrics at step 11700
INFO:tensorflow:
                        + DetectionBoxes_Precision/mAP: 0.313534
I0109 22:59:20.835347 140110343554496 model_lib_v2.py:1010]
                                                                 + DetectionBoxes_Precision/mAP: 0.313534
INFO:tensorflow:
                        + DetectionBoxes_Precision/mAP@.50IOU: 0.642573
                                                                 + DetectionBoxes_Precision/mAP@.50IOU: 0.642573
I0109 22:59:20.836146 140110343554496 model_lib_v2.py:1010]
INFO:tensorflow:
                        + DetectionBoxes Precision/mAP@.75IOU: 0.182703
I0109 22:59:20.836694 140110343554496 model_lib_v2.py:1010]
                                                                 + DetectionBoxes_Precision/mAP@.75IOU: 0.182703
INFO:tensorflow:
                        + DetectionBoxes Precision/mAP (small): 0.000000
I0109 22:59:20.837198 140110343554496 model lib v2.pv:1010]
                                                                 + DetectionBoxes Precision/mAP (small): 0.000000
INFO:tensorflow:
                        + DetectionBoxes Precision/mAP (medium): 0.200446
I0109 22:59:20.837694 140110343554496 model_lib_v2.py:1010]
                                                                 + DetectionBoxes_Precision/mAP (medium): 0.200446
INFO:tensorflow:
                        + DetectionBoxes Precision/mAP (large):
                                                                0.342099
I0109 22:59:20.838170 140110343554496 model_lib_v2.py:1010]
                                                                 + DetectionBoxes_Precision/mAP (large): 0.342099
INFO:tensorflow:
                        + DetectionBoxes_Recall/AR@1: 0.200263
I0109 22:59:20.838645 140110343554496 model_lib_v2.py:1010]
                                                                 + DetectionBoxes_Recall/AR@1: 0.200263
INFO:tensorflow:
                        + DetectionBoxes_Recall/AR@10: 0.396053
I0109 22:59:20.839138 140110343554496 model lib v2.py:1010]
                                                                   DetectionBoxes Recall/AR@10: 0.396053
                        + DetectionBoxes_Recall/AR@100: 0.437807
INFO:tensorflow:
I0109 22:59:20.839626 140110343554496 model_lib_v2.py:1010]
                                                                  DetectionBoxes_Recall/AR@100: 0.437807
INFO:tensorflow:
                        + DetectionBoxes_Recall/AR@100 (small)
                                                                 0.000000
I0109 22:59:20.840081 140110343554496 model_lib_v2.py:1010]
                                                                 + DetectionBoxes_Recall/AR@100 (small): 0.000000
INFO:tensorflow:
                        + DetectionBoxes_Recall/AR@100 (medium): 0.331250
I0109 22:59:20.840557 140110343554496 model_lib_v2.py:1010]
                                                                 + DetectionBoxes_Recall/AR@100 (medium): 0.331250
                        + DetectionBoxes Recall/AR@100 (large)
INFO:tensorflow:
                                                                 0.462894
I0109 22:59:20.841067 140110343554496 model_lib_v2.py:1010]
                                                                 + DetectionBoxes Recall/AR@100 (large): 0.462894
                        + Loss/localization_loss: 0.296767
INFO:tensorflow:
I0109 22:59:20.841483 140110343554496 model_lib_v2.py:1010]
                                                                 + Loss/localization_loss: 0.296767
                        + Loss/classification loss: 0.510632
INFO:tensorflow:
I0109 22:59:20.841907 140110343554496 model lib v2.py:1010]
                                                                 + Loss/classification loss: 0.510632
INFO:tensorflow:
                        + Loss/regularization loss: 0.542540
I0109 22:59:20.842327 140110343554496 model_lib_v2.py:1010]
                                                                 + Loss/regularization_loss: 0.542540
INFO:tensorflow:
                        + Loss/total_loss: 1.349939
I0109 22:59:20.842776 140110343554496 model_lib_v2.py:1010]
                                                                 + Loss/total_loss: 1.349939
```



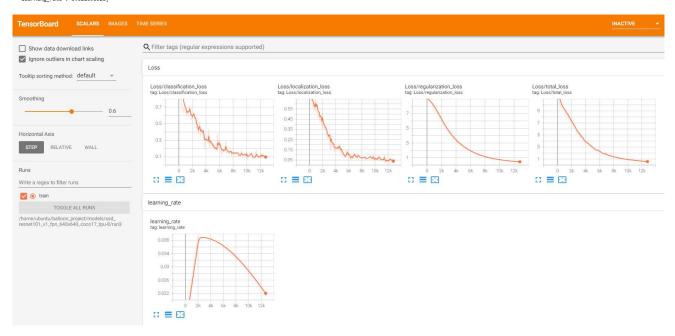




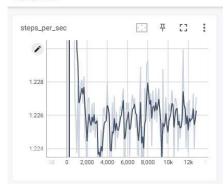


Section 3.3: Run 3 Detailed Results

```
'Loss/localization_loss': 0.03190211,
'Loss/regularization_loss': 0.43443272,
'Loss/total_loss': 0.54976285,
'learning_rate': 0.021096328}
```



steps_per_sec



```
Average Precision (AP) @[ IoU=0.50:0.95 |
Average Precision (AP) @[ IoU=0.50 |
Average Precision (AP) @[ IoU=0.50 |
Average Precision (AP) @[ IoU=0.50:0.95 |
Average Precision (AP) @[ IoU=0.50:0.95 |
Average Precision (AP) @[ IoU=0.50:0.95 |
Average Recall (AR) @[ IoU=0.50:0.95 |
                                                                                                    maxDets=100 ] = 0.309
maxDets=100 ] = 0.604
maxDets=100 ] = 0.313
maxDets=100 ] = 0.000
maxDets=100 ] = 0.189
                                                                                          all
                                                                            area=
                                                                            area= all
area= small
                                                                             area=medium
                                                                            area=medium
area= large
area= all
area= all
area= small
                                                                                                                             = 0.323
                                                                                                     maxDets=100
                                                                                                     maxDets= 1
maxDets= 10
                                                                                                    maxDets= 1 ] = 0.210
maxDets= 10 ] = 0.398
maxDets=100 ] = 0.443
maxDets=100 ] = 0.000
maxDets=100 ] = 0.327
maxDets=100 ] = 0.459
                                                                            area=medium
area= large
 INFO:tensorflow:Eval metrics at step 12900
+ DetectionBoxes Precision/mAP@.50IOU: 0.604386
                                                                                                                   DetectionBoxes_Precision/mAP@.75IOU: 0.312803
+ DetectionBoxes Precision/mAP (small): 0.000000
                                                                                                                  DetectionBoxes_Precision/mAP (medium): 0.188822
                                                                                                                + DetectionBoxes Precision/mAP (large): 0.323309
                                                                                                                   DetectionBoxes Recall/AR@1: 0.210088
                                                                                                                  DetectionBoxes Recall/AR@10: 0.398421
                                                                                                                  DetectionBoxes_Recall/AR@100: 0.442719
INFO:tensorflow: + DetectionBoxes_Recall/AR@100 (smail):
10109 22:55:50.125130 139659940021696 model_lib_v2.py:1010]
INFO:tensorflow: + DetectionBoxes_Recall/AR@100 (medium)
10109 22:55:50.125130 139659940021696 model_lib_v2.py:1010]
INFO:tensorflow: + DetectionBoxes_Recall/AR@100 (large):
                                                                                                                + DetectionBoxes Recall/AR@100 (small): 0.000000
                                                                                                                 0.327206
                                                                                                                  DetectionBoxes_Recall/AR@100 (medium): 0.327206
+ DetectionBoxes Recall/AR@100 (large): 0.459369
                                                                                                                + Loss/localization loss: 0.308534
                                                                                                                + Loss/classification_loss: 0.566454
                                                                                                                + Loss/regularization_loss: 0.434367
 INFO:tensorflow:
                                         + Loss/total loss: 1.309355
I0109 22:55:50.127390 139659940021696 model_lib_v2.py:1010]
                                                                                                               + Loss/total loss: 1.309355
```

Section 3.4: Run 4 Detailed Results

```
I0110 02:06:22.563429 140644024722880 model_lib_v2.py:701] {'Loss/classification_loss': 0.12946908,
'loss/localization_loss': 0.040291637,
'Loss/calization_loss': 0.040291637,
'Loss/regularization_loss': 0.401158,
'Loss/total_loss': 0.57091874,
'learning_rate': 0.022681937}
INFO:tensorflow:Step 12400 per-step time 0.785s
I0110 02:07:41.039009 140644024722880 model lib_v2.py:698] Step 12400 per-step time 0.785s INFO:tensorflow:{'Loss/classification_loss': 0.06427159,
'Loss/localization loss': 0.028724974,
   'Loss/regularization_loss': 0.028724974,
   'Loss/regularization_loss': 0.39531687,
   'Loss/total_loss': 0.48831344,
   'learning_rate': 0.022418866}

I0110 02:07:41.039325 140644024722880 model_lib_v2.py:701] {'Loss/classification_loss': 0.06427159,
    'Loss/localization_loss': 0.028724974,
'Loss/regularization_loss': 0.39531687,
'Loss/total_loss': 0.48831344,
'learning_rate': 0.022418866}
INFO:tensorflow:Step 12500 per-step time 0.785s
I0110 02:08:59.592703 140644024722880 model_lib_v2.py:698] Step 12500 per-step time 0.785s INFO:tensorflow:{\loss/classification_loss': 0.099045835,
    'Loss/localization_loss': 0.029932687,
'Loss/regularization_loss': 0.38945168,
'Loss/total_loss': 0.51843023,
'learning_rate': 0.022155251}
I0110 02:08:59.502973 140644024722880 model_lib_v2.py:701] {'Loss/classification_loss': 0.099045835,
Total 02:08:59.5029/3 140044024/22800 model_int_

'Loss/localization_loss': 0.38945168,

'Loss/total_loss': 0.51843023,

'Learning_rate': 0.022155251}

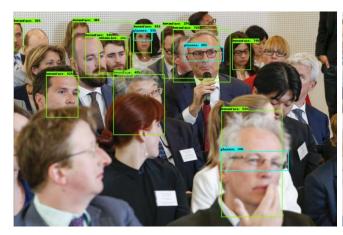
INFO:tensorflow:Step 12600 per-step time 0.785s
I0110 02:10:17.982205 140644024722880 model_lib_v2.py:698] Step 12600 per-step time 0.785s INFO:tensorflow:{'Loss/classification_loss': 0.120541014,
    'Loss/localization_loss': 0.0406868,
'Loss/regularization_loss': 0.3838333,
'Loss/total_loss': 0.5450611,
'learning_rate': 0.02189114}
I0110 02:10:17.982475 140644024722880 model_lib_v2.py:701] {'Loss/classification_loss': 0.120541014,
IO110 02:10:17.982475 140644024722880 model_lib_v2.py:701] {'Loss/classification_loss': 0.:
'Loss/regularization_loss': 0.0406868,
'Loss/regularization_loss': 0.3838333,
'Loss/total_loss': 0.5450611,
'learning_rate': 0.02189114}
INFO:tensorflow:Step 12700 per-step time 0.785s
IO110 02:11:36.527723 140644024722880 model_lib_v2.py:698] Step 12700 per-step time 0.785s
INFO:tensorflow:{'Loss/classification_loss': 0.08524129,
'Loss/Classification_loss': 0.08524129,
INFO:TensorTion: { Loss/classification_loss : 0.08524129, 
    'Loss/localization_loss': 0.041681, 
    'Loss/regularization_loss': 0.37828946, 
    'Loss/total_loss': 0.5052117, 
    'learning_rate': 0.201626579} 
IO110 02:11:36.527987 140644024722880 model_lib_v2.py:701] { 'Loss/classification_loss': 0.08524129,
   'Loss/localization_loss': 0.041681,
'Loss/regularization_loss': 0.37828946,
'Loss/total_loss': 0.5052117,
'learning_rate': 0.021626579}
```



Section 4: Test Image and Test Video

Using the trained Run 2 model that had the best performance metric, object detection for the two classes was done on an out-of-sample test image and test video.

Below shows the results from the object detection model on the test image, comparing a minimum threshold of 0.25 versus 0.50.





Below shows the a few frames of results from the object detection model on the test video. The predicted bounding boxes do not appear consistently throughout the video, where at quite a few intervals there is no output detection. We also note that the pretrained model was trained on the COCO dataset, where one of the objects is "Person". Our "Human face" object in could be said to have an overlap with the COCO "Person" object the model was pretrained on. In contrast, "Glasses" is a completely new type of object. This could explain why the object detection model is better able to detect "Human face" over "Glasses".

In addition, we note that 80 train images and 20 validation images was used was for the experimental runs. Lifting time constraints and increasing the number of annotated train and validation images will allow the model to "see" and be trained on many more examples of our selected two object types, and make the model more robust, possibly leading to an increase in the mAP@IoU=0.50 performance metric





