## Autonomous\_driving\_application\_Car\_detection\_2022\_05\_04\_09\_36\_33-Copy1

May 4, 2022

```
[1]: import argparse
     import os
     import matplotlib.pyplot as plt
     from matplotlib.pyplot import imshow
     import scipy.io
     import scipy.misc
     import numpy as np
     import pandas as pd
     import PIL
     from PIL import ImageFont, ImageDraw, Image
     import tensorflow as tf
     from tensorflow.python.framework.ops import EagerTensor
     from tensorflow.keras.models import load_model
     from yad2k.models.keras_yolo import yolo_head
     from yad2k.utils.utils import draw_boxes, get_colors_for_classes, scale_boxes,_
     →read_classes, read_anchors, preprocess_image
     %matplotlib inline
```

```
[2]: # UNQ_C1 (UNIQUE CELL IDENTIFIER, DO NOT EDIT)
# GRADED FUNCTION: yolo_filter_boxes

def yolo_filter_boxes(boxes, box_confidence, box_class_probs, threshold = .6):
    box_scores = box_confidence * box_class_probs
    box_classes = tf.math.argmax(box_scores, axis = -1)
    box_class_scores = tf.math.reduce_max(box_scores, axis = -1)

filtering_mask = box_class_scores >= threshold

scores = tf.boolean_mask(box_class_scores, filtering_mask)
    boxes = tf.boolean_mask(boxes, filtering_mask)
```

```
classes = tf.boolean_mask(box_classes, filtering_mask)
return scores, boxes, classes
```

```
[3]: # BEGIN UNIT TEST
     tf.random.set seed(10)
     box_confidence = tf.random.normal([19, 19, 5, 1], mean=1, stddev=4, seed = 1)
     boxes = tf.random.normal([19, 19, 5, 4], mean=1, stddev=4, seed = 1)
     box_class_probs = tf.random.normal([19, 19, 5, 80], mean=1, stddev=4, seed = 1)
     scores, boxes, classes = yolo_filter_boxes(boxes, box_confidence,_
     →box_class_probs, threshold = 0.5)
     print("scores[2] = " + str(scores[2].numpy()))
     print("boxes[2] = " + str(boxes[2].numpy()))
     print("classes[2] = " + str(classes[2].numpy()))
     print("scores.shape = " + str(scores.shape))
     print("boxes.shape = " + str(boxes.shape))
     print("classes.shape = " + str(classes.shape))
     assert type(scores) == EagerTensor, "Use tensorflow functions"
     assert type(boxes) == EagerTensor, "Use tensorflow functions"
     assert type(classes) == EagerTensor, "Use tensorflow functions"
     assert scores.shape == (1789,), "Wrong shape in scores"
     assert boxes.shape == (1789, 4), "Wrong shape in boxes"
     assert classes.shape == (1789,), "Wrong shape in classes"
     assert np.isclose(scores[2].numpy(), 9.270486), "Values are wrong on scores"
     assert np.allclose(boxes[2].numpy(), [4.6399336, 3.2303846, 4.431282, -2.
     →202031]), "Values are wrong on boxes"
     assert classes[2].numpy() == 8, "Values are wrong on classes"
     print("\033[92m All tests passed!")
     # END UNIT TEST
    scores[2] = 9.270486
    boxes[2] = [ 4.6399336  3.2303846  4.431282  -2.202031 ]
    classes[2] = 8
    scores.shape = (1789,)
    boxes.shape = (1789, 4)
    classes.shape = (1789,)
     All tests passed!
[6]: # UNQ C2 (UNIQUE CELL IDENTIFIER, DO NOT EDIT)
     # GRADED FUNCTION: iou
     def iou(box1, box2):
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(box1_x1, box1_y1, box1_x2, box1_y2) = box1
(box2_x1, box2_y1, box2_x2, box2_y2) = box2
xi1 = \max(box1[0], box2[0])
yi1 = max(box1[1],box2[1])
xi2 = min(box1[2],box2[2])
yi2 = min(box1[3],box2[3])
inter_width = max(xi2 - xi1,0)
inter_height = max(yi2 - yi1,0)
inter_area = inter_width*inter_height
box1_area = (box1[3] - box1[1])*(box1[2] - box1[0])
box2_area = (box2[3] - box2[1])*(box2[2] - box2[0])
union_area = box1_area + box2_area - inter_area
# compute the IoU
iou = inter_area/union_area
### END CODE HERE
return iou
```

```
[7]: # BEGIN UNIT TEST
     ## Test case 1: boxes intersect
     box1 = (2, 1, 4, 3)
     box2 = (1, 2, 3, 4)
     print("iou for intersecting boxes = " + str(iou(box1, box2)))
     assert iou(box1, box2) < 1, "The intersection area must be always smaller or ⊔
     →equal than the union area."
     assert np.isclose(iou(box1, box2), 0.14285714), "Wrong value. Check your
     →implementation. Problem with intersecting boxes"
     ## Test case 2: boxes do not intersect
     box1 = (1,2,3,4)
     box2 = (5,6,7,8)
     print("iou for non-intersecting boxes = " + str(iou(box1,box2)))
     assert iou(box1, box2) == 0, "Intersection must be 0"
     ## Test case 3: boxes intersect at vertices only
     box1 = (1,1,2,2)
     box2 = (2,2,3,3)
     print("iou for boxes that only touch at vertices = " + str(iou(box1,box2)))
     assert iou(box1, box2) == 0, "Intersection at vertices must be 0"
     ## Test case 4: boxes intersect at edge only
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```
box1 = (1,1,3,3)
      box2 = (2,3,3,4)
      print("iou for boxes that only touch at edges = " + str(iou(box1,box2)))
      assert iou(box1, box2) == 0, "Intersection at edges must be 0"
      print("\033[92m All tests passed!")
      # END UNIT TEST
     iou for intersecting boxes = 0.14285714285714285
     iou for non-intersecting boxes = 0.0
     iou for boxes that only touch at vertices = 0.0
     iou for boxes that only touch at edges = 0.0
      All tests passed!
[12]: # UNQ_C3 (UNIQUE CELL IDENTIFIER, DO NOT EDIT)
      # GRADED FUNCTION: yolo non max suppression
      def yolo non max suppression(scores, boxes, classes, max_boxes = 10,__
       →iou_threshold = 0.5):
          max boxes tensor = tf.Variable(max boxes, dtype='int32') # tensor to be |
       →used in tf.image.non_max_suppression()
          nms_indices = tf.image.
       →non_max_suppression(boxes,scores,max_boxes_tensor,iou_threshold =__
       →iou_threshold)
          scores = tf.gather(scores,nms_indices)
          boxes = tf.gather(boxes,nms indices)
          classes = tf.gather(classes,nms_indices)
          return scores, boxes, classes
[13]: # BEGIN UNIT TEST
      tf.random.set seed(10)
      scores = tf.random.normal([54,], mean=1, stddev=4, seed = 1)
      boxes = tf.random.normal([54, 4], mean=1, stddev=4, seed = 1)
      classes = tf.random.normal([54,], mean=1, stddev=4, seed = 1)
      scores, boxes, classes = yolo_non_max_suppression(scores, boxes, classes)
      assert type(scores) == EagerTensor, "Use tensoflow functions"
      print("scores[2] = " + str(scores[2].numpy()))
      print("boxes[2] = " + str(boxes[2].numpy()))
      print("classes[2] = " + str(classes[2].numpy()))
      print("scores.shape = " + str(scores.numpy().shape))
      print("boxes.shape = " + str(boxes.numpy().shape))
```

print("classes.shape = " + str(classes.numpy().shape))

```
assert type(scores) == EagerTensor, "Use tensoflow functions"
      assert type(boxes) == EagerTensor, "Use tensoflow functions"
      assert type(classes) == EagerTensor, "Use tensoflow functions"
      assert scores.shape == (10,), "Wrong shape"
      assert boxes.shape == (10, 4), "Wrong shape"
      assert classes.shape == (10,), "Wrong shape"
      assert np.isclose(scores[2].numpy(), 8.147684), "Wrong value on scores"
      assert np.allclose(boxes[2].numpy(), [ 6.0797963, 3.743308, 1.3914018, -0.
      \hookrightarrow34089637]), "Wrong value on boxes"
      assert np.isclose(classes[2].numpy(), 1.7079165), "Wrong value on classes"
      print("\033[92m All tests passed!")
      # END UNIT TEST
     scores[2] = 8.147684
     boxes[2] = [6.0797963]
                              3.743308 1.3914018 -0.34089637]
     classes[2] = 1.7079165
     scores.shape = (10,)
     boxes.shape = (10, 4)
     classes.shape = (10,)
      All tests passed!
[14]: def yolo_boxes_to_corners(box_xy, box_wh):
          """Convert YOLO box predictions to bounding box corners."""
          box_mins = box_xy - (box_wh / 2.)
          box_maxes = box_xy + (box_wh / 2.)
          return tf.keras.backend.concatenate([
              box_mins[..., 1:2], \# y_min
              box_mins[..., 0:1], \# x_min
              box_maxes[..., 1:2], \# y_max
              box_maxes[..., 0:1] # x_max
          ])
[21]: # UNQ_C4 (UNIQUE CELL IDENTIFIER, DO NOT EDIT)
      # GRADED FUNCTION: yolo_eval
      def yolo_eval(yolo_outputs, image_shape = (720, 1280), max_boxes=10,__
       ⇒score_threshold=.6, iou_threshold=.5):
          # Retrieve outputs of the YOLO model (1 line)
          box_xy, box_wh, box_confidence, box_class_probs = yolo_outputs
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# Convert boxes to be ready for filtering functions (convert boxes box_xy_\_\)

and box_wh to corner coordinates)
boxes = yolo_boxes_to_corners(box_xy, box_wh)

# Use one of the functions you've implemented to perform Score-filtering_\]

with a threshold of score_threshold (1 line)
scores, boxes, classes = yolo_filter_boxes(boxes, box_confidence,_\]

box_class_probs, threshold = score_threshold)

# Scale boxes back to original image shape.
boxes = scale_boxes(boxes, image_shape)

# Use one of the functions you've implemented to perform Non-max_\]

suppression with a threshold of iou_threshold (1 line)
scores, boxes, classes = yolo_non_max_suppression(scores, boxes, classes,_\]

max_boxes = max_boxes, iou_threshold = iou_threshold)

### END CODE HERE ###

return scores, boxes, classes
```

```
[22]: # BEGIN UNIT TEST
      tf.random.set_seed(10)
      yolo outputs = (tf.random.normal([19, 19, 5, 2], mean=1, stddev=4, seed = 1),
                      tf.random.normal([19, 19, 5, 2], mean=1, stddev=4, seed = 1),
                      tf.random.normal([19, 19, 5, 1], mean=1, stddev=4, seed = 1),
                      tf.random.normal([19, 19, 5, 80], mean=1, stddev=4, seed = 1))
      scores, boxes, classes = yolo_eval(yolo_outputs)
      print("scores[2] = " + str(scores[2].numpy()))
      print("boxes[2] = " + str(boxes[2].numpy()))
      print("classes[2] = " + str(classes[2].numpy()))
      print("scores.shape = " + str(scores.numpy().shape))
      print("boxes.shape = " + str(boxes.numpy().shape))
      print("classes.shape = " + str(classes.numpy().shape))
      assert type(scores) == EagerTensor, "Use tensoflow functions"
      assert type(boxes) == EagerTensor, "Use tensoflow functions"
      assert type(classes) == EagerTensor, "Use tensoflow functions"
      assert scores.shape == (10,), "Wrong shape"
      assert boxes.shape == (10, 4), "Wrong shape"
      assert classes.shape == (10,), "Wrong shape"
      assert np.isclose(scores[2].numpy(), 171.60194), "Wrong value on scores"
      assert np.allclose(boxes[2].numpy(), [-1240.3483, -3212.5881, -645.78, 2024.
      \hookrightarrow3052]), "Wrong value on boxes"
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assert np.isclose(classes[2].numpy(), 16), "Wrong value on classes"
     print("\033[92m All tests passed!")
     # END UNIT TEST
    scores[2] = 171.60194
    boxes[2] = [-1240.3483 - 3212.5881 - 645.78 2024.3052]
    classes[2] = 16
    scores.shape = (10,)
    boxes.shape = (10, 4)
    classes.shape = (10,)
     All tests passed!
[9]: class names = read classes("model data/coco classes.txt")
     anchors = read_anchors("model_data/yolo_anchors.txt")
     model image size = (608, 608) # Same as yolo model input layer size
[18]: yolo_model = load_model("model_data/" ,compile=True)
    WARNING:tensorflow: No training configuration found in save file, so the model
    was *not* compiled. Compile it manually.
[19]: volo model.summary()
    Model: "functional_1"
                                  Output Shape Param # Connected to
    Layer (type)
                                 [(None, 608, 608, 3) 0
    input_1 (InputLayer)
     conv2d (Conv2D)
                                 (None, 608, 608, 32) 864
     ______
    batch_normalization (BatchNorma (None, 608, 608, 32) 128 conv2d[0][0]
    leaky_re_lu (LeakyReLU)
                                  (None, 608, 608, 32) 0
    batch_normalization[0][0]
    max_pooling2d (MaxPooling2D) (None, 304, 304, 32) 0
    leaky_re_lu[0][0]
                                 (None, 304, 304, 64) 18432
    conv2d_1 (Conv2D)
```

max_pooling2d[0][0]						
batch_normalization_1 (BatchNor						
leaky_re_lu_1 (LeakyReLU) batch_normalization_1[0][0]	(None,	304,	304,	64)	0	
max_pooling2d_1 (MaxPooling2D) leaky_re_lu_1[0][0]						
 conv2d_2 (Conv2D) max_pooling2d_1[0][0]	(None,					
batch_normalization_2 (BatchNor						
leaky_re_lu_2 (LeakyReLU) batch_normalization_2[0][0]	(None,					
 conv2d_3 (Conv2D) leaky_re_lu_2[0][0]	(None,	152,	152,	64)	8192	
batch_normalization_3 (BatchNor	(None,	152,	152,	64)	256	conv2d_3[0][0]
leaky_re_lu_3 (LeakyReLU) batch_normalization_3[0][0]	(None,				0	
 conv2d_4 (Conv2D) leaky_re_lu_3[0][0]	(None,					
batch_normalization_4 (BatchNor	(None,	152,	152,	128	512	conv2d_4[0][0]
leaky_re_lu_4 (LeakyReLU) batch_normalization_4[0][0]	(None,	152,	152,	128	0	
max_pooling2d_2 (MaxPooling2D)	(None,					

leaky_re_lu_4[0][0]						
conv2d_5 (Conv2D) max_pooling2d_2[0][0]	(None,	76,	76,	256)	294912	
batch_normalization_5 (BatchNor						
leaky_re_lu_5 (LeakyReLU) batch_normalization_5[0][0]	(None,	76,	76,	256)	0	
conv2d_6 (Conv2D) leaky_re_lu_5[0][0]	(None,	-	-			
batch_normalization_6 (BatchNor						
leaky_re_lu_6 (LeakyReLU) batch_normalization_6[0][0]	(None,	76,			0	
conv2d_7 (Conv2D) leaky_re_lu_6[0][0]	(None,	76,			294912	
batch_normalization_7 (BatchNor	(None,	76,	76,	256)	1024	conv2d_7[0][0]
leaky_re_lu_7 (LeakyReLU) batch_normalization_7[0][0]	(None,				0	
	(None,	38,	38,	256)	0	
conv2d_8 (Conv2D) max_pooling2d_3[0][0]	(None,	38,	38,	512)	1179648	
batch_normalization_8 (BatchNor	(None,	38,	38,	512)	2048	conv2d_8[0][0]
leaky_re_lu_8 (LeakyReLU)	(None,					

batch_normalization_8[0][0]						
conv2d_9 (Conv2D) leaky_re_lu_8[0][0]	(None,	38,	38,	256)	131072	
batch_normalization_9 (BatchNor						_
leaky_re_lu_9 (LeakyReLU) batch_normalization_9[0][0]	(None,	38,	38,	256)	0	
conv2d_10 (Conv2D) leaky_re_lu_9[0][0]					1179648	
batch_normalization_10 (BatchNo						
leaky_re_lu_10 (LeakyReLU) batch_normalization_10[0][0]	(None,	38,	38,	512)	0	
conv2d_11 (Conv2D) leaky_re_lu_10[0][0]	(None,	38,	38,	256)	131072	
batch_normalization_11 (BatchNo	(None,	38,	38,	256)	1024	conv2d_11[0][0]
leaky_re_lu_11 (LeakyReLU) batch_normalization_11[0][0]	(None,	38,				
conv2d_12 (Conv2D) leaky_re_lu_11[0][0]			38,	512)	1179648	
batch_normalization_12 (BatchNo	(None,	38,	38,	512)	2048	conv2d_12[0][0]
leaky_re_lu_12 (LeakyReLU) batch_normalization_12[0][0]	(None,	38,	38,	512)	0	
	(None,					==== <b>-3=</b>

leaky_re_lu_12[0][0]						
conv2d_13 (Conv2D) max_pooling2d_4[0][0]	(None,	19,	19,	1024)	4718592	
batch_normalization_13 (BatchNo						
leaky_re_lu_13 (LeakyReLU) batch_normalization_13[0][0]	(None,					
conv2d_14 (Conv2D) leaky_re_lu_13[0][0]					524288	
batch_normalization_14 (BatchNo						
leaky_re_lu_14 (LeakyReLU) batch_normalization_14[0][0]	(None,	19,	19,	512)	0	
conv2d_15 (Conv2D) leaky_re_lu_14[0][0]	(None,	19,	19,	1024)	4718592	
batch_normalization_15 (BatchNo	(None,	19,	19,	1024)	4096	conv2d_15[0][0]
leaky_re_lu_15 (LeakyReLU) batch_normalization_15[0][0]	(None,					
conv2d_16 (Conv2D) leaky_re_lu_15[0][0]		19,	19,	512)	524288	
batch_normalization_16 (BatchNo	(None,	19,	19,	512)	2048	conv2d_16[0][0]
leaky_re_lu_16 (LeakyReLU) batch_normalization_16[0][0]	(None,	19,	19,	512)	0	
conv2d_17 (Conv2D)					4718592	<b>_</b>

leaky_re_lu_16[0][0]						
batch_normalization_17 (BatchNo	(None,	19,	19,	1024)	4096	conv2d_17[0][0]
leaky_re_lu_17 (LeakyReLU) batch_normalization_17[0][0]	(None,	19,	19,	1024)	0	
 conv2d_18 (Conv2D) leaky_re_lu_17[0][0]					9437184	
batch_normalization_18 (BatchNo	(None,	19,	19,	1024)	4096	conv2d_18[0][0]
conv2d_20 (Conv2D) leaky_re_lu_12[0][0]	(None,	38,	38,	64)	32768	
leaky_re_lu_18 (LeakyReLU) batch_normalization_18[0][0]	(None,	19,	19,	1024)	0	
batch_normalization_20 (BatchNo	(None,	38,	38,	64)	256	
conv2d_19 (Conv2D) leaky_re_lu_18[0][0]					9437184	
leaky_re_lu_20 (LeakyReLU) batch_normalization_20[0][0]						
batch_normalization_19 (BatchNo	(None,	19,	19,	1024)	4096	conv2d_19[0][0]
space_to_depth_x2 (Lambda) leaky_re_lu_20[0][0]	(None,	19,	19,	256)	0	
leaky_re_lu_19 (LeakyReLU) batch_normalization_19[0][0]	(None,	19,	19,	1024)	0	
concatenate (Concatenate)	(None,					<b></b>

```
space_to_depth_x2[0][0]
    leaky_re_lu_19[0][0]
    conv2d 21 (Conv2D)
                              (None, 19, 19, 1024) 11796480
    concatenate[0][0]
    ______
    batch_normalization_21 (BatchNo (None, 19, 19, 1024) 4096 conv2d_21[0][0]
    leaky_re_lu_21 (LeakyReLU)
                            (None, 19, 19, 1024) 0
    batch_normalization_21[0][0]
           ______
    conv2d_22 (Conv2D)
                             (None, 19, 19, 425) 435625
    leaky_re_lu_21[0][0]
    ______
    ===========
    Total params: 50,983,561
    Trainable params: 50,962,889
    Non-trainable params: 20,672
    ______
    _____
[23]: def predict(image_file):
        # Preprocess your image
        image, image_data = preprocess_image("images/" + image_file,__
     \rightarrowmodel_image_size = (608, 608))
        yolo_model_outputs = yolo_model(image_data)
        yolo_outputs = yolo_head(yolo_model_outputs, anchors, len(class_names))
        out_scores, out_boxes, out_classes = yolo_eval(yolo_outputs, [image.
     \rightarrowsize[1], image.size[0]], 10, 0.3, 0.5)
        # Print predictions info
        print('Found {} boxes for {}'.format(len(out_boxes), "images/" +__
     →image_file))
        # Generate colors for drawing bounding boxes.
        colors = get_colors_for_classes(len(class_names))
        # Draw bounding boxes on the image file
        #draw boxes2(image, out scores, out boxes, out classes, class names,
     →colors, image_shape)
        draw_boxes(image, out_boxes, out_classes, class_names, out_scores)
        # Save the predicted bounding box on the image
```

```
image.save(os.path.join("out", image_file), quality=100)
# Display the results in the notebook
output_image = Image.open(os.path.join("out", image_file))
imshow(output_image)
return out_scores, out_boxes, out_classes
```

[27]: out\_scores, out\_boxes, out\_classes = predict("Transpo-Mumbaitraffic-1079622456.

→ jpg")

```
Found 10 boxes for images/Transpo-Mumbaitraffic-1079622456.jpg person 0.71 (1310, 394) (1431, 798) person 0.68 (1127, 444) (1268, 907) person 0.67 (1672, 493) (1818, 970) person 0.67 (974, 379) (1087, 841) person 0.67 (753, 311) (900, 736) person 0.66 (828, 1012) (956, 1200) person 0.64 (2089, 1015) (2273, 1200) person 0.63 (1879, 495) (2034, 937) person 0.63 (906, 400) (1046, 837) car 0.62 (1050, 1063) (1527, 1189)
```

