Keras-YOLOv3

April 4, 2020

1 Part1: Build YOLOv3

```
[1]: # create a YOLOv3 Keras model and save it to file

# based on https://github.com/experiencor/keras-yolo3
import struct
import numpy as np
from keras.layers import Conv2D
from keras.layers import Input
from keras.layers import BatchNormalization
from keras.layers import LeakyReLU
from keras.layers import ZeroPadding2D
from keras.layers import UpSampling2D
from keras.layers.merge import add, concatenate
from keras.models import Model
```

Using TensorFlow backend.

```
[2]: def _conv_block(inp, convs, skip=True):
         111
         inp: input
         convs: array of sets of configurations
         111
         x = inp
         count = 0
         for conv in convs:
             if count == (len(convs) - 2) and skip:
                 skip\_connection = x
             count += 1
             if conv['stride'] > 1:
                 x = ZeroPadding2D(((1,0),(1,0)))(x) # peculiar padding as darknet_{\sqcup}
      \rightarrowprefer left and top
             x = Conv2D(conv['filter'],
                         conv['kernel'],
                         strides = conv['stride'],
                         padding = 'valid' if conv['stride'] > 1 else 'same', #_
      →peculiar padding as darknet prefer left and top
                                 = 'conv_' + str(conv['layer_idx']),
                         name
```

```
use_bias =False if conv['bnorm'] else True)(x)
       if conv['bnorm']:
           x = BatchNormalization(epsilon=0.001, name='bnorm_' +__

→str(conv['layer_idx']))(x)
       if conv['leaky']:
           x = LeakyReLU(alpha=0.1, name='leaky ' + str(conv['layer idx']))(x)
   return add([skip_connection, x]) if skip else x
def make_yolov3_model():
   111
   yolo_82 = layer81
   yolo_94 = layer84 + skip61 + Upsampling + concatenate + _conv_block5
   yolo_106 = layer96 + skip36 + Upsampling + concatenate + _conv_block7
   input_image = Input(shape=(None, None, 3))
   # Layer 0 => 4
   x = _conv_block(input_image, [{'filter': 32, 'kernel': 3, 'stride': 1,__
{'filter': 64, 'kernel': 3, 'stride': 2, __
{'filter': 32, 'kernel': 1, 'stride': 1, |
{'filter': 64, 'kernel': 3, 'stride': 1,⊔
→ 'bnorm': True, 'leaky': True, 'layer_idx': 3}])
   # Layer 5 => 8
   x = _conv_block(x, [{'filter': 128, 'kernel': 3, 'stride': 2, 'bnorm': __
→True, 'leaky': True, 'layer_idx': 5},
                      {'filter': 64, 'kernel': 1, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 6},
                      {'filter': 128, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 7}])
   # Layer 9 => 11
   x = _conv_block(x, [{'filter': 64, 'kernel': 1, 'stride': 1, 'bnorm': __
→True, 'leaky': True, 'layer_idx': 9},
                      {'filter': 128, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 10}])
   # Layer 12 => 15
   x = _conv_block(x, [{'filter': 256, 'kernel': 3, 'stride': 2, 'bnorm': __
→True, 'leaky': True, 'layer_idx': 12},
                      {'filter': 128, 'kernel': 1, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 13},
                      {'filter': 256, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 14}])
   # Layer 16 => 36
   for i in range(7):
```

```
x = _conv_block(x, [{'filter': 128, 'kernel': 1, 'stride': 1, 'bnorm': __
→True, 'leaky': True, 'layer_idx': 16+i*3},
                          {'filter': 256, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 17+i*3}])
   skip_36 = x
   # Layer 37 => 40
   x = _conv_block(x, [{'filter': 512, 'kernel': 3, 'stride': 2, 'bnorm': __
→True, 'leaky': True, 'layer_idx': 37},
                      {'filter': 256, 'kernel': 1, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 38},
                      {'filter': 512, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 39}])
   # Layer 41 => 61
  for i in range(7):
      x = _conv_block(x, [{'filter': 256, 'kernel': 1, 'stride': 1, 'bnorm':u
→True, 'leaky': True, 'layer_idx': 41+i*3},
                          {'filter': 512, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 42+i*3}])
   skip 61 = x
   # Layer 62 => 65
   x = _conv_block(x, [{'filter': 1024, 'kernel': 3, 'stride': 2, 'bnorm':u
→True, 'leaky': True, 'layer_idx': 62},
                      {'filter': 512, 'kernel': 1, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 63},
                      {'filter': 1024, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 64}])
   # Layer 66 => 74
   for i in range(3):
      x = _conv_block(x, [{'filter': 512, 'kernel': 1, 'stride': 1, 'bnorm': u
→True, 'leaky': True, 'layer_idx': 66+i*3},
                          {'filter': 1024, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 67+i*3}])
   # Layer 75 => 79
   x = _conv_block(x, [{'filter': 512, 'kernel': 1, 'stride': 1, 'bnorm': __
→True, 'leaky': True, 'layer_idx': 75},
                      {'filter': 1024, 'kernel': 3, 'stride': 1, 'bnorm': 11
→True, 'leaky': True, 'layer_idx': 76},
                      {'filter': 512, 'kernel': 1, 'stride': 1, 'bnorm': u
→True, 'leaky': True, 'layer_idx': 77},
                      {'filter': 1024, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 78},
                      {'filter': 512, 'kernel': 1, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 79}], skip=False)
   # Layer 80 => 82
   yolo_82 = _conv_block(x, [{'filter': 1024, 'kernel': 3, 'stride': 1,__
```

```
{'filter': 255, 'kernel': 1, 'stride': 1, |
→'bnorm': False, 'leaky': False, 'layer_idx': 81}], skip=False)
  # Layer 83 => 86
  x = conv block(x, [{'filter': 256, 'kernel': 1, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 84}], skip=False)
  x = UpSampling2D(2)(x)
  x = concatenate([x, skip_61])
  # Layer 87 => 91
  x = _conv_block(x, [{'filter': 256, 'kernel': 1, 'stride': 1, 'bnorm': u
→True, 'leaky': True, 'layer_idx': 87},
                     {'filter': 512, 'kernel': 3, 'stride': 1, 'bnorm': u
→True, 'leaky': True, 'layer_idx': 88},
                     {'filter': 256, 'kernel': 1, 'stride': 1, 'bnorm': u
→True, 'leaky': True, 'layer_idx': 89},
                     {'filter': 512, 'kernel': 3, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 90},
                     {'filter': 256, 'kernel': 1, 'stride': 1, 'bnorm':
→True, 'leaky': True, 'layer_idx': 91}], skip=False)
  # Layer 92 => 94
  yolo_94 = _conv_block(x, [{'filter': 512, 'kernel': 3, 'stride': 1, 'bnorm':
→ True, 'leaky': True, 'layer_idx': 92},
                           {'filter': 255, 'kernel': 1, 'stride': 1, 'bnorm':
→ False, 'leaky': False, 'layer_idx': 93}], skip=False)
  # Layer 95 => 98
  x = _conv_block(x, [{'filter': 128, 'kernel': 1, 'stride': 1, 'bnorm': __
→True, 'leaky': True, 'layer_idx': 96}], skip=False)
  x = UpSampling2D(2)(x)
  x = concatenate([x, skip_36])
  # Layer 99 => 106
  yolo_106 = _conv_block(x, [{'filter': 128, 'kernel': 1, 'stride': 1, _
{'filter': 256, 'kernel': 3, 'stride': 1, _
                'leaky': True, 'layer_idx': 100},
{'filter': 128, 'kernel': 1, 'stride': 1, |
                'leaky': True, 'layer_idx': 101},
{'filter': 256, 'kernel': 3, 'stride': 1,,,
                'leaky': True, 'layer_idx': 102},
{'filter': 128, 'kernel': 1, 'stride': 1, |
               'leaky': True, 'layer_idx': 103},
{'filter': 256, 'kernel': 3, 'stride': 1, __
→'bnorm': True, 'leaky': True, 'layer_idx': 104},
                            {'filter': 255, 'kernel': 1, 'stride': 1, __
→ 'bnorm': False, 'leaky': False, 'layer_idx': 105}], skip=False)
  model = Model(input_image, [yolo_82, yolo_94, yolo_106])
  return model
```

```
class WeightReader:
   def __init__(self, weight_file):
       with open(weight_file, 'rb') as w_f:
            # i: int
                    = struct.unpack('i', w_f.read(4))
            major,
                   = struct.unpack('i', w_f.read(4))
            minor,
            revision, = struct.unpack('i', w_f.read(4))
            if (major*10 + minor) >= 2 and major < 1000 and minor < 1000:
                w f.read(8)
            else:
                w f.read(4)
            transpose = (major > 1000) or (minor > 1000)
            binary = w f.read()
        self.offset = 0
        self.all_weights = np.frombuffer(binary, dtype='float32')
   def read_bytes(self, size):
        self.offset = self.offset + size
        return self.all_weights[self.offset-size:self.offset]
   def load_weights(self, model):
        for i in range(106):
            try:
                conv layer = model.get layer('conv ' + str(i))
                print("loading weights of convolution #" + str(i))
                if i not in [81, 93, 105]:
                    norm_layer = model.get_layer('bnorm_' + str(i))
                    size = np.prod(norm_layer.get_weights()[0].shape)
                    beta = self.read_bytes(size) # bias
                    gamma = self.read_bytes(size) # scale
                    mean = self.read_bytes(size) # mean
                          = self.read_bytes(size) # variance
                    weights = norm_layer.set_weights([gamma, beta, mean, var])
                if len(conv_layer.get_weights()) > 1:
                    bias = self.read_bytes(np.prod(conv_layer.
→get_weights()[1].shape))
                    kernel = self.read_bytes(np.prod(conv_layer.
 →get_weights()[0].shape))
                    kernel = kernel.reshape(list(reversed(conv_layer.
 →get_weights()[0].shape)))
                    kernel = kernel.transpose([2,3,1,0])
                    conv_layer.set_weights([kernel, bias])
                else:
                    kernel = self.read_bytes(np.prod(conv_layer.
→get_weights()[0].shape))
                    kernel = kernel.reshape(list(reversed(conv_layer.

    get_weights()[0].shape)))
```

```
[3]: # define the model
model = make_yolov3_model()
# load the model weights
weight_reader = WeightReader('yolov3.weights')
# set the model weights into the model
weight_reader.load_weights(model)
# save the model to file
model.save('model.h5')
```

```
loading weights of convolution #0
loading weights of convolution #1
loading weights of convolution #2
loading weights of convolution #3
no convolution #4
loading weights of convolution #5
loading weights of convolution #6
loading weights of convolution #7
no convolution #8
loading weights of convolution #9
loading weights of convolution #10
no convolution #11
loading weights of convolution #12
loading weights of convolution #13
loading weights of convolution #14
no convolution #15
loading weights of convolution #16
loading weights of convolution #17
no convolution #18
loading weights of convolution #19
loading weights of convolution #20
no convolution #21
loading weights of convolution #22
loading weights of convolution #23
no convolution #24
loading weights of convolution #25
loading weights of convolution #26
no convolution #27
loading weights of convolution #28
loading weights of convolution #29
```

no convolution #30		
loading weights of	convolution	#31
loading weights of	convolution	#32
no convolution #33		
loading weights of	convolution	#34
loading weights of	convolution	#35
no convolution #36		
loading weights of	convolution	#37
loading weights of	convolution	#38
loading weights of	convolution	#39
no convolution #40		
loading weights of	convolution	#41
loading weights of	convolution	#42
no convolution #43		
loading weights of	convolution	#44
loading weights of	convolution	#45
no convolution #46		
loading weights of	convolution	#47
loading weights of	convolution	#48
no convolution #49		
loading weights of	convolution	#50
loading weights of	convolution	#51
no convolution #52		
loading weights of	convolution	#53
loading weights of	convolution	#54
no convolution #55		
loading weights of	convolution	#56
loading weights of	convolution	#57
no convolution #58		
loading weights of	convolution	#59
loading weights of	convolution	#60
no convolution #61		
loading weights of	convolution	#62
loading weights of	convolution	#63
loading weights of	convolution	#64
no convolution #65		
loading weights of	convolution	#66
loading weights of	convolution	#67
no convolution #68	0011101401011	
loading weights of	convolution	#69
loading weights of	convolution	
no convolution #71	0011101401011	
loading weights of	convolution	#72
loading weights of	convolution	#73
no convolution #74		
loading weights of	convolution	#75
loading weights of	convolution	#76
loading weights of	convolution	
	2311.01401011	

```
loading weights of convolution #78
loading weights of convolution #79
loading weights of convolution #80
loading weights of convolution #81
no convolution #82
no convolution #83
loading weights of convolution #84
no convolution #85
no convolution #86
loading weights of convolution #87
loading weights of convolution #88
loading weights of convolution #89
loading weights of convolution #90
loading weights of convolution #91
loading weights of convolution #92
loading weights of convolution #93
no convolution #94
no convolution #95
loading weights of convolution #96
no convolution #97
no convolution #98
loading weights of convolution #99
loading weights of convolution #100
loading weights of convolution #101
loading weights of convolution #102
loading weights of convolution #103
loading weights of convolution #104
loading weights of convolution #105
```

2 Part2: Make a prediction

```
[4]: # load yolov3 model and perform object detection
# based on https://github.com/experiencor/keras-yolo3
from numpy import expand_dims
from keras.models import load_model
from keras.preprocessing.image import load_img
from keras.preprocessing.image import img_to_array
```

```
[5]: # load and prepare an image
def load_image_pixels(filename, shape):
    # load the image to get its shape
    image = load_img(filename)
    width, height = image.size
    # load the image with the required size
    image = load_img(filename, target_size=shape)
    # convert to numpy array
    image = img_to_array(image)
```

```
# scale pixel values to [0, 1]
image = image.astype('float32')
image /= 255.0
# add a dimension so that we have one sample
image = expand_dims(image, 0)
return image, width, height
```

```
[6]: # load yolov3 model
  model = load_model('model.h5')
  # define the expected input shape for the model
  input_w, input_h = 416, 416
  # define our new photo
  photo_filename = 'img/zebra.jpg'
  # load and prepare image
  image, image_w, image_h = load_image_pixels(photo_filename, (input_w, input_h))
  # make prediction
  yhat = model.predict(image)
  # summarize the shape of the list of arrays
  print([a.shape for a in yhat])
```

/Users/chanho/miniconda3/envs/kd/lib/python3.5/sitepackages/keras/engine/saving.py:269: UserWarning: No training configuration
found in save file: the model was *not* compiled. Compile it manually.
warnings.warn('No training configuration found in save file: '
[(1, 13, 13, 255), (1, 26, 26, 255), (1, 52, 52, 255)]

3 Part3: Make a Prediction and Interpret Result

```
[7]: # load yolov3 model and perform object detection
# based on https://github.com/experiencor/keras-yolo3
import numpy as np
from numpy import expand_dims
from keras.models import load_model
from keras.preprocessing.image import load_img
from keras.preprocessing.image import img_to_array
from matplotlib import pyplot
from matplotlib.patches import Rectangle
```

```
class BoundBox:
    def __init__(self, xmin, ymin, xmax, ymax, objness = None, classes = None):
        self.xmin = xmin
        self.ymin = ymin
        self.xmax = xmax
        self.ymax = ymax
        self.objness = objness
        self.classes = classes
```

```
self.label = -1
        self.score = -1
    def get_label(self):
        if self.label == -1:
            self.label = np.argmax(self.classes)
        return self.label
    def get_score(self):
        if self.score == -1:
            self.score = self.classes[self.get_label()]
        return self.score
def _sigmoid(x):
    return 1. / (1. + np.exp(-x))
# [(1, 13, 13, 255), (1, 26, 26, 255), (1, 52, 52, 255)]
def decode_netout(netout, anchors, obj_thresh, net_h, net_w):
    grid_h, grid_w = netout.shape[:2] # (13, 13)
    nb box = 3
    netout = netout.reshape((grid_h, grid_w, nb_box, -1)) # (13, 13, 3, 1)
    nb class = netout.shape[-1] - 5 # 85 - 5 = 80, this is the number of labels
    boxes = []
    # : means as many as needed
    netout[..., :2] = _sigmoid(netout[..., :2])
    netout[..., 4:] = \_sigmoid(netout[..., 4:])
    # np.newaxis: increase the dimension
    netout[..., 5:] = netout[..., 4][..., np.newaxis] * netout[..., 5:]
    netout[..., 5:] *= netout[..., 5:] > obj_thresh
    for i in range(grid_h*grid_w):
        row = i / grid_w
        col = i % grid_w
        for b in range(nb_box):
            # 4th element is objectness score
            objectness = netout[int(row)][int(col)][b][4]
            if(objectness.all() <= obj_thresh):</pre>
                continue
            # first 4 elements are x, y, w, and h
            x, y, w, h = netout[int(row)][int(col)][b][:4]
            x = (col + x) / grid_w # center position, unit: image width
            y = (row + y) / grid_h # center position, unit: image height
            w = anchors[2 * b + 0] * np.exp(w) / net_w # unit: image width
            h = anchors[2 * b + 1] * np.exp(h) / net_h # unit: image height
            # last elements are class probabilities
```

```
classes = netout[int(row)][col][b][5:]
            box = BoundBox(x-w/2, y-h/2, x+w/2, y+h/2, objectness, classes)
            boxes.append(box)
    return boxes
def correct_yolo_boxes(boxes, image_h, image_w, net_h, net_w):
    new_w, new_h = net_w, net_h
    for i in range(len(boxes)):
        x_offset, x_scale = (net_w - new_w)/2./net_w, float(new_w)/net_w
        y_offset, y_scale = (net_h - new_h)/2./net_h, float(new_h)/net_h
        boxes[i].xmin = int((boxes[i].xmin - x_offset) / x_scale * image_w)
        boxes[i].xmax = int((boxes[i].xmax - x_offset) / x_scale * image_w)
        boxes[i].ymin = int((boxes[i].ymin - y_offset) / y_scale * image_h)
        boxes[i].ymax = int((boxes[i].ymax - y_offset) / y_scale * image_h)
def _interval_overlap(interval_a, interval_b):
    x1, x2 = interval_a
    x3, x4 = interval_b
    if x3 < x1:
        if x4 < x1:
            return 0
        else:
            return min(x2,x4) - x1
    else:
        if x2 < x3:
            return 0
        else:
            return min(x2,x4) - x3
def bbox_iou(box1, box2):
    intersect_w = _interval_overlap([box1.xmin, box1.xmax], [box2.xmin, box2.
\rightarrowxmax])
    intersect_h = _interval_overlap([box1.ymin, box1.ymax], [box2.ymin, box2.
\rightarrowymax])
    intersect = intersect_w * intersect_h
    w1, h1 = box1.xmax-box1.xmin, box1.ymax-box1.ymin
    w2, h2 = box2.xmax-box2.xmin, box2.ymax-box2.ymin
    union = w1*h1 + w2*h2 - intersect
    return float(intersect) / union
def do_nms(boxes, nms_thresh):
    if len(boxes) > 0:
        nb_class = len(boxes[0].classes)
    else:
        return
    for c in range(nb_class):
        sorted_indices = np.argsort([-box.classes[c] for box in boxes])
```

```
for i in range(len(sorted_indices)):
            index_i = sorted_indices[i]
            if boxes[index_i].classes[c] == 0:
                continue
            for j in range(i+1, len(sorted_indices)):
                index_j = sorted_indices[j]
                if bbox_iou(boxes[index_i], boxes[index_j]) >= nms_thresh:
                    boxes[index_j].classes[c] = 0
# load and prepare an image
def load image pixels(filename, shape):
   # load the image to get its shape
   image = load_img(filename)
   width, height = image.size
    # load the image with the required size
   image = load_img(filename, target_size=shape)
    # convert to numpy array
   image = img_to_array(image)
   # scale pixel values to [0, 1]
   image = image.astype('float32')
   image /= 255.0
    # add a dimension so that we have one sample
   image = expand_dims(image, 0)
   return image, width, height
# get all of the results above a threshold
def get_boxes(boxes, labels, thresh):
   v_boxes, v_labels, v_scores = list(), list(), list()
    # enumerate all boxes
   for box in boxes:
        # enumerate all possible labels
        for i in range(len(labels)):
            # check if the threshold for this label is high enough
            if box.classes[i] > thresh:
                v_boxes.append(box)
                v_labels.append(labels[i])
                v scores.append(box.classes[i]*100)
                # don't break, many labels may trigger for one box
   return v_boxes, v_labels, v_scores
# draw all results
def draw_boxes(filename, v_boxes, v_labels, v_scores):
    # load the image
   data = pyplot.imread(filename)
    # plot the image
   pyplot.imshow(data)
    # get the context for drawing boxes
```

```
ax = pyplot.gca()
# plot each box
for i in range(len(v_boxes)):
   box = v_boxes[i]
    # get coordinates
   y1, x1, y2, x2 = box.ymin, box.xmin, box.ymax, box.xmax
    # calculate width and height of the box
   width, height = x2 - x1, y2 - y1
    # create the shape
   rect = Rectangle((x1, y1), width, height, fill=False, color='white')
    # draw the box
   ax.add_patch(rect)
    # draw text and score in top left corner
   label = "%s (%.3f)" % (v_labels[i], v_scores[i])
   pyplot.text(x1, y1, label, color='white')
# show the plot
pyplot.show()
```

```
[10]: # load yolov3 model
      model = load_model('model.h5')
      # define the expected input shape for the model
      input_w, input_h = 416, 416
      # define our new photo
      photo_filename = 'img/zebra.jpg'
      # load and prepare image
      image, image w, image h = load image pixels(photo filename, (input w, input h))
      # make prediction
      yhat = model.predict(image)
      # summarize the shape of the list of arrays
      print([a.shape for a in yhat])
      # define the anchors
      anchors = [[116,90, 156,198, 373,326], [30,61, 62,45, 59,119], [10,13, 16,30],
      <u>→</u>33,23]]
      # define the probability threshold for detected objects
      class threshold = 0.6
      boxes = list()
      for i in range(len(yhat)):
          # decode the output of the network
          boxes += decode_netout(yhat[i][0], anchors[i], class_threshold, input_h,_u
      →input_w)
      # correct the sizes of the bounding boxes for the shape of the image
      correct_yolo_boxes(boxes, image_h, image_w, input_h, input_w)
      # suppress non-maximal boxes
      do_nms(boxes, 0.5)
      # define the labels
      labels = ["person", "bicycle", "car", "motorbike", "aeroplane", "bus", "train",
```

```
"boat", "traffic light", "fire hydrant", "stop sign", "parking meter", __

→ "bench",

   "bird", "cat", "dog", "horse", "sheep", "cow", "elephant", "bear", "zebra",
"backpack", "umbrella", "handbag", "tie", "suitcase", "frisbee", "skis", [

→ "snowboard",

   "sports ball", "kite", "baseball bat", "baseball glove", "skateboard",
"tennis racket", "bottle", "wine glass", "cup", "fork", "knife", "spoon", [

→"bowl", "banana",
   "apple", "sandwich", "orange", "broccoli", "carrot", "hot dog", "pizza", "
"chair", "sofa", "pottedplant", "bed", "diningtable", "toilet", "
"remote", "keyboard", "cell phone", "microwave", "oven", "toaster", "sink", [
"book", "clock", "vase", "scissors", "teddy bear", "hair drier", "
→"toothbrush"]
# get the details of the detected objects
v_boxes, v_labels, v_scores = get_boxes(boxes, labels, class_threshold)
# summarize what we found
for i in range(len(v_boxes)):
   print(v_labels[i], v_scores[i])
# draw what we found
draw_boxes(photo_filename, v_boxes, v_labels, v_scores)
```

```
/Users/chanho/miniconda3/envs/kd/lib/python3.5/site-packages/keras/engine/saving.py:269: UserWarning: No training configuration found in save file: the model was *not* compiled. Compile it manually. warnings.warn('No training configuration found in save file: '
[(1, 13, 13, 255), (1, 26, 26, 255), (1, 52, 52, 255)]
zebra 94.91060376167297
zebra 99.86329674720764
zebra 96.8708872795105
```

