Ex No: 8

OBJECT DETECTION WITH YOLO3

Aim:

To build an object detection model with YOLO3 using Keras/TensorFlow.

Procedure:

- 1. Download and load the dataset.
- 2. Perform analysis and preprocessing of the dataset.
- 3. Build a simple neural network model using Keras/TensorFlow.
- 4. Compile and fit the model.
- 5. Perform prediction with the test dataset.
- 6. Calculate performance metrics.

Program:

```
# 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))
def decode_netout(netout, anchors, obj_thresh, net_h, net_w):
       grid_h, grid_w = netout.shape[:2]
       nb_box = 3
       netout = netout.reshape((grid_h, grid_w, nb_box, -1))
       nb_class = netout.shape[-1] - 5
```

```
boxes = [] netout[..., :2] = _sigmoid(netout[..., :2]) netout[...,
       4:] = _sigmoid(netout[..., 4:]) 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): 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)):
```

```
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.xmax])
       intersect_h = _interval_overlap([box1.ymin, box1.ymax], [box2.ymin, box2.ymax])
       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
```

x_offset, x_scale = (net_w - new_w)/2./net_w, float(new_w)/net_w

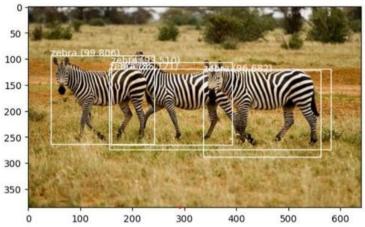
```
return float(intersect) / union
def do_nms(boxes, nms_thresh):
       if len(boxes) > 0:
               nb_class = len(boxes[0].classes)
       else:
               return
       for c in range(nb_class):
               sorted_indices = np.argsort([-box.classes[c] for box in boxes])
               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()
# 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 = '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, 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, 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", "truck",
       "boat", "traffic light", "fire hydrant", "stop sign", "parking meter", "bench",
       "bird", "cat", "dog", "horse", "sheep", "cow", "elephant", "bear", "zebra", "giraffe",
       "backpack", "umbrella", "handbag", "tie", "suitcase", "frisbee", "skis", "snowboard",
       "sports ball", "kite", "baseball bat", "baseball glove", "skateboard", "surfboard",
       "tennis racket", "bottle", "wine glass", "cup", "fork", "knife", "spoon", "bowl", "banana",
       "apple", "sandwich", "orange", "broccoli", "carrot", "hot dog", "pizza", "donut", "cake",
       "chair", "sofa", "pottedplant", "bed", "diningtable", "toilet", "tymonitor", "laptop",
"mouse",
       "remote", "keyboard", "cell phone", "microwave", "oven", "toaster", "sink",
"refrigerator",
       "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
```

Output:



Result:

Object Detection using YOLO has been successfully implemented.