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)):
```

```
x offset, x scale = (\text{net } \text{w - new } \text{w})/2./\text{net } \text{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 x^2 < x^3:
                       return 0
               else:
                      return min(x2,x4) - x3
def bbox iou(box1, box2):
       intersect w = interval \text{ overlap}([box1.xmin, box1.xmax], [box2.xmin, box2.xmax])
       intersect h = interval \text{ 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
```

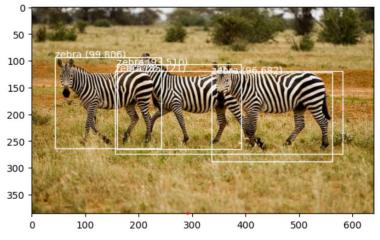
```
return float(intersect) / union
def do_nms(boxes, nms_thresh):
       if len(boxes) > 0:
              nb class = len(boxes[0].classes)
       else:
              return
       for c in range(nb class):
               sorted indices = np.argsort([-box.classes[c] for box in boxes])
              for 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 = x^2 - x^1, y^2 - y^1
              # 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", "tvmonitor", "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.