#### 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 = (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:
```

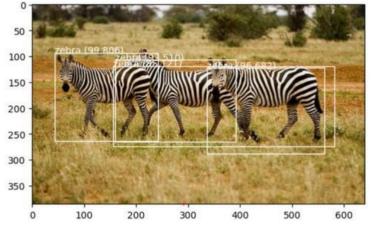
```
if x4 < x1:
                      return 0
              else:
                      return min(x2,x4) - x1
       else:
              if x^2 < x^3:
                      return 0
              else:
                      return min(x2,x4) - x3
def bbox_iou(box1, box2):
       intersect_w = _interval_overlap([box1.xmin, 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
       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_volo_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
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)
```

# Output:



## Result:

Object Detection using YOLO has been successfully implemented.