DETECTION WITH YOLO3

Ex No: 8 Aim:

OBJECT

To build an object detection model with YOLO3 using

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

class BoundBox:

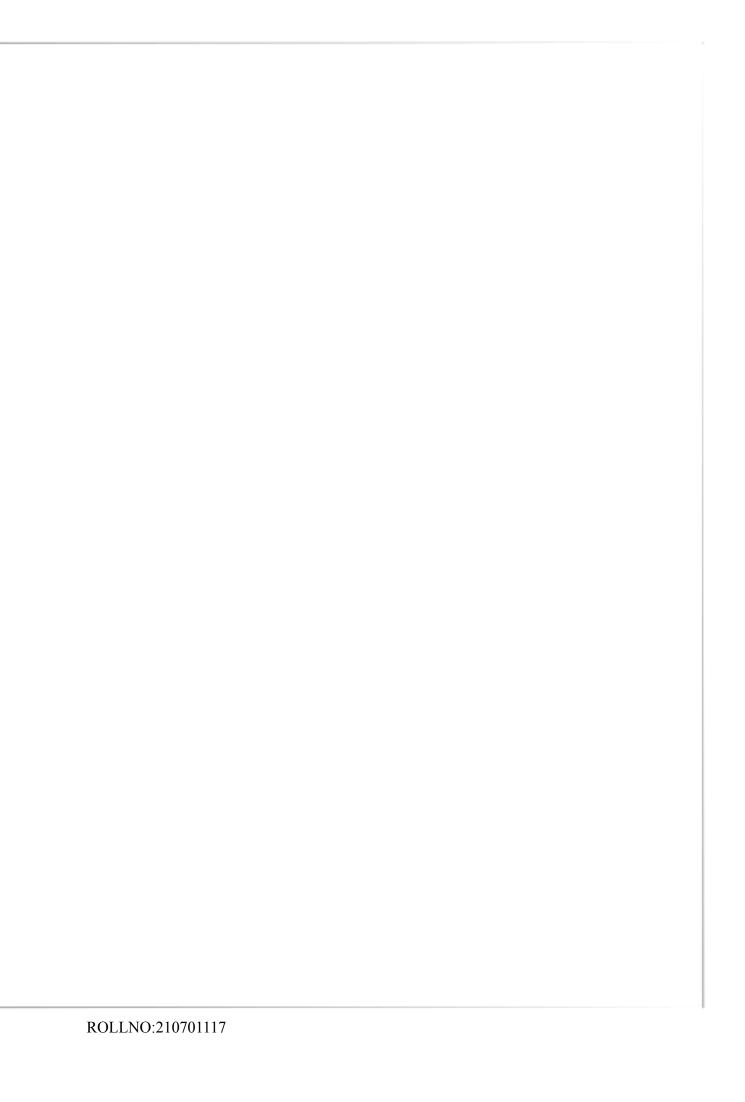
def_init_(self, xmin, ymin, xmax, ymax, objness = None, classes =

None): self.xmin = xmin

from matplotlib.patches import Rectangle

```
ROLLNO:210701128 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[..., :2]
       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)):
```

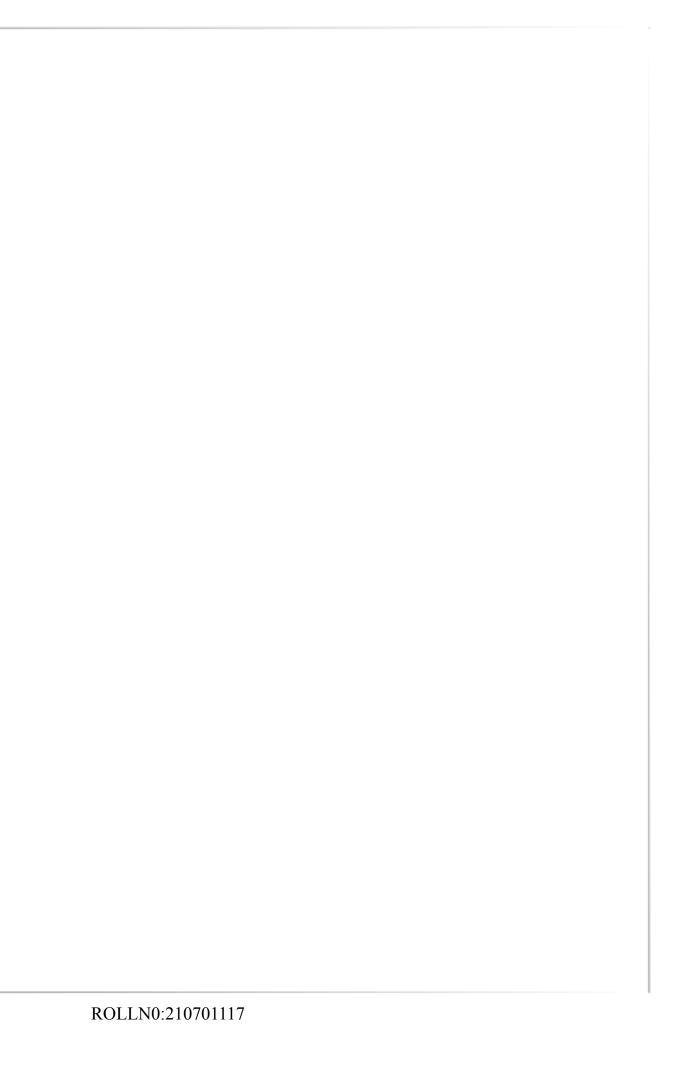


```
y offset, y scale = (net h - new h)/2./net h, float(new h)/net h
          boxes[i].xmin = int((boxes[i].xmin - x offset) / x scale * image w)
         boxes[i].xmax = int((boxes[i].xmax - x offset) / x scale * image w)
             boxes[i].ymin = int((boxes[i].ymin - y offset) / y scale * image 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 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, box1.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
```

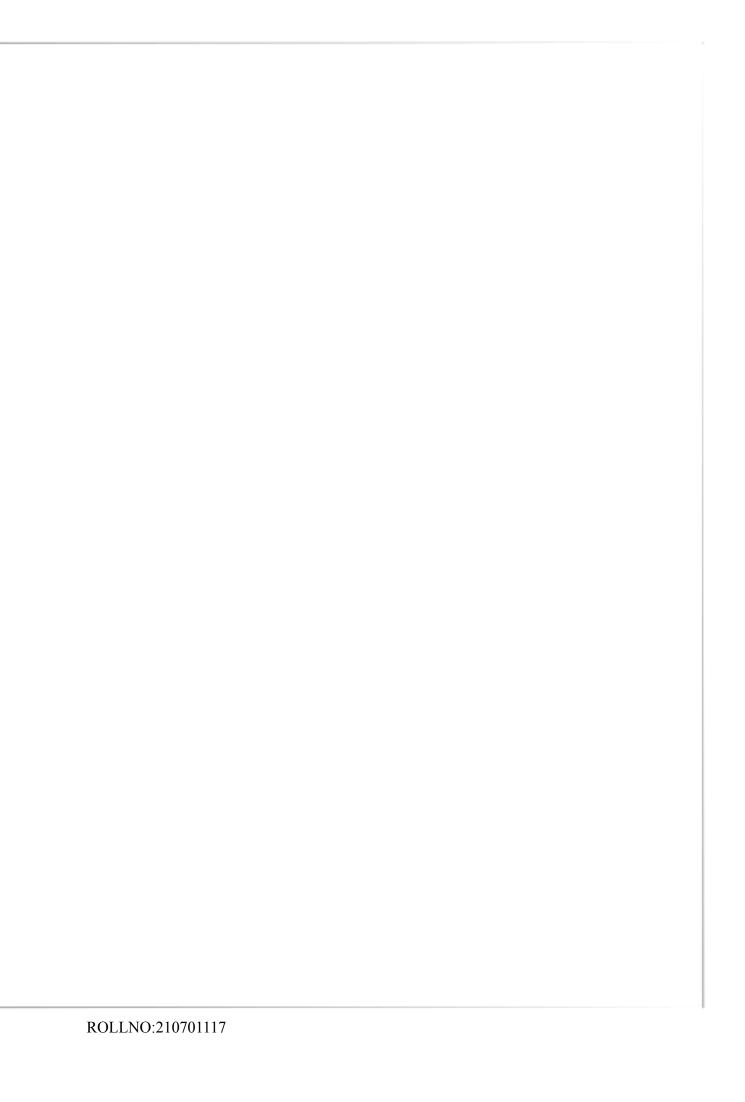
x offset, x scale = (net w - new w)/2./net w, float(new w)/net w



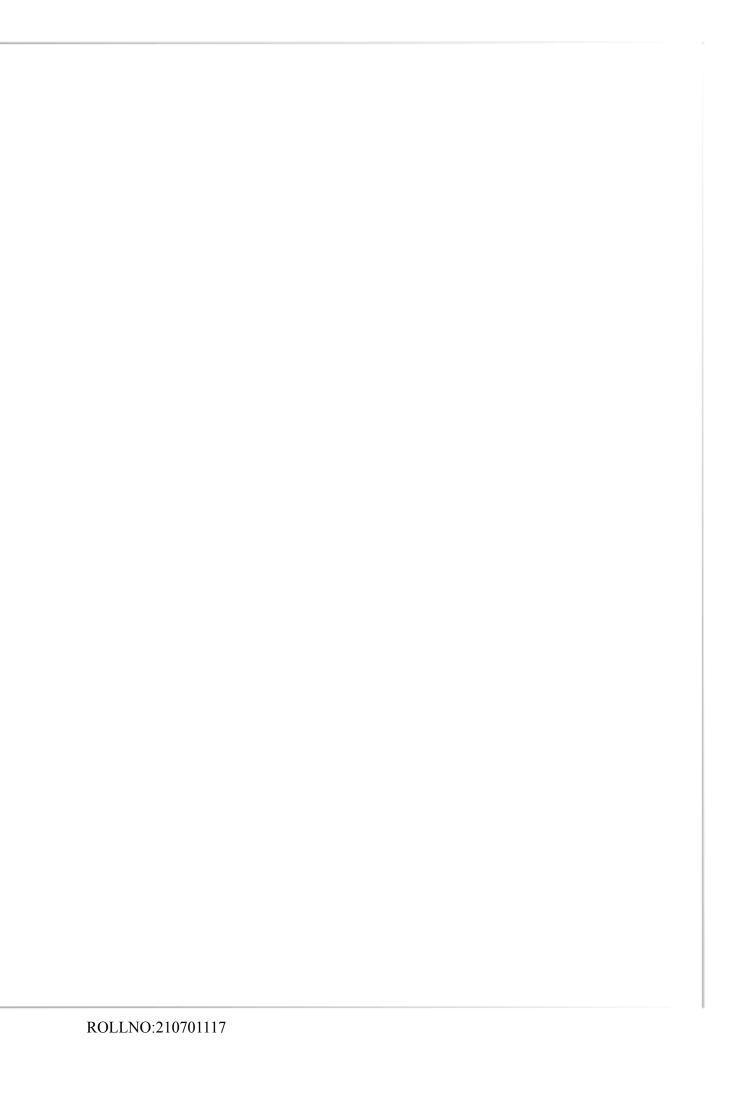
```
return float(intersect) / union
def do nms(boxes, nms thresh):
       if len(boxes) > 0:
              nb class = len(boxes[0].classes)
       else:
              return
       for c in range(nb class):
              sorted indices = np.argsort([-box.classes[c] for box in boxes])
              for 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()
```

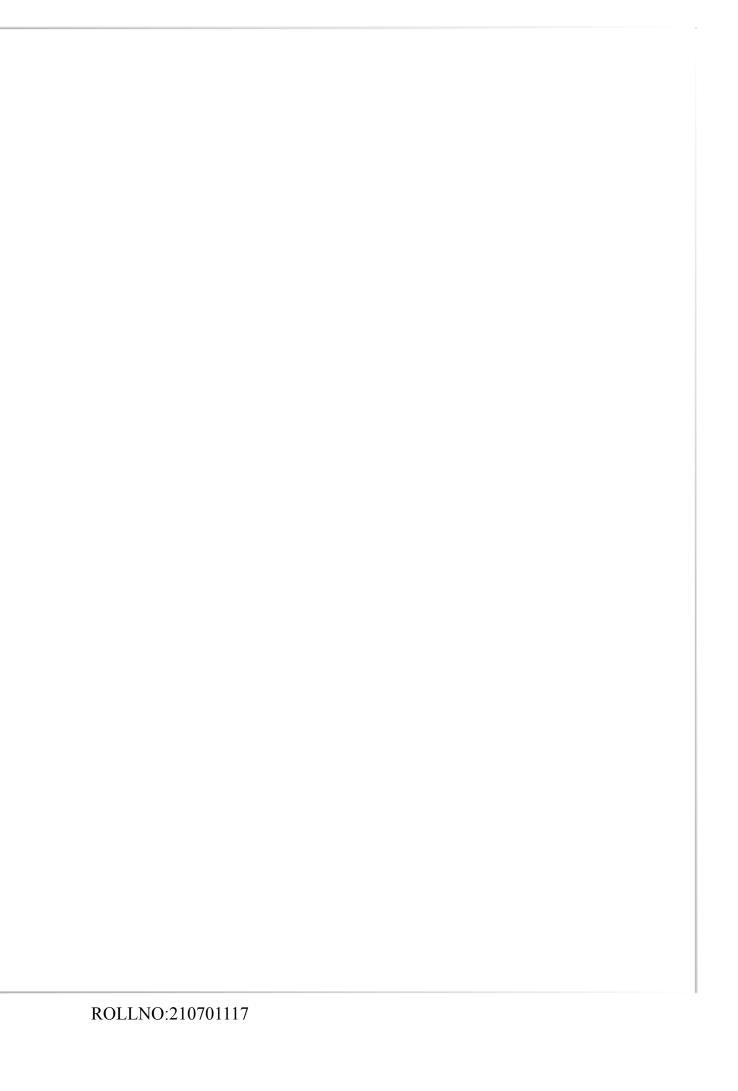


```
# plot each box
      for i in range(len(v boxes)):
             box = v_boxes[i]
             # get coordinates
             y1, x1, y2, x2 = box.ymin, box.xmin, box.ymax, box.xmax
             # calculate width and height of the box
             width, height = x^2 - x^1, y^2 - y^1
             # 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
```

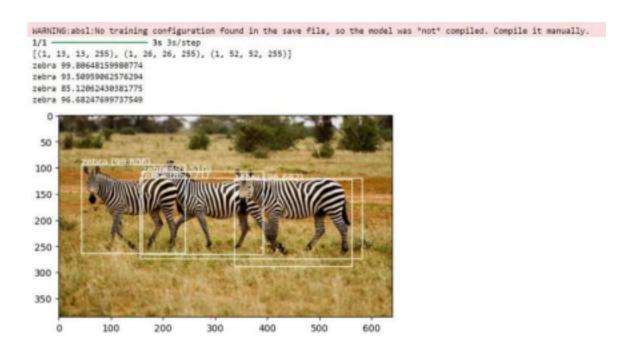


```
print([a.shape for a in yhat])
# define the anchors
anchors = [[116,90, 156,198, 373,326], [30,61, 62,45, 59,119], [10,13,
16,30, 33,23]] # define the probability threshold for detected objects
class threshold = 0.6
boxes = list()
for i in range(len(yhat)):
      # decode the output of the network
      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", "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.