

## Yolo3 pytorch Object detection

```
from __future__ import division

from models import *
from utils.utils import *
from utils.datasets import *

import os
import sys
import time
import datetime
import argparse

from PIL import Image

import torch
from torch.utils.data import DataLoader
from torchvision import datasets
from torch.autograd import Variable

import matplotlib.pyplot as plt
import matplotlib.patches as patches
from matplotlib.ticker import NullLocator

device = torch.device("cuda" if torch.cuda.is_available() else "cpu")

os.makedirs("output", exist_ok=True)

# Set up model
model = Darknet("config/yolov3.cfg", img_size=416).to(device)
# Load darknet weights
model.load_darknet_weights("weights/yolov3.weights")

model.eval() # Set in evaluation mode

dataloader = DataLoader(
    ImageFolder("data/samples", img_size=416),
    batch_size=1,
    shuffle=False,
    num_workers=0,
)

classes = load_classes("data/coco.names") # Extracts class labels from file

Tensor = torch.cuda.FloatTensor if torch.cuda.is_available() else torch.FloatTensor
```

image = r1 + classes + image paths

[https://colab.research.google.com/github/nicewook/datascience\\_exercise/blob/master/PyTorch\\_YOLOv3.ipynb#scrollTo=AGNQ8gywrzDW&printMode...](https://colab.research.google.com/github/nicewook/datascience_exercise/blob/master/PyTorch_YOLOv3.ipynb#scrollTo=AGNQ8gywrzDW&printMode...) 1/6

```
imgs = [] # Scores image paths
img_detections = [] # Stores detections for each image index

print("\nPerforming object detection:")
prev_time = time.time()
for batch_i, (img_paths, input_imgs) in enumerate(dataloader):
    # Configure input
    input_imgs = Variable(input_imgs.type(Tensor))

    # Get detections
    with torch.no_grad():
        detections = model(input_imgs)
        detections = non_max_suppression(detections, 0.8, 0.4)

    # Log progress
    current_time = time.time()
    inference_time = datetime.timedelta(seconds=current_time - prev_time)
    prev_time = current_time
    print("\t+ Batch %d, Inference Time: %s" % (batch_i, inference_time))

    # Save image and detections
    imgs.extend(img_paths)
    img_detections.extend(detections)

# Bounding-box colors
cmap = plt.get_cmap("tab20b")
colors = [cmap(i) for i in np.linspace(0, 1, 20)]

print("\nSaving images:")
# Iterate through images and save plot of detections
for img_i, (path, detections) in enumerate(zip(imgs, img_detections)):

    print("(%) Image: '%s'" % (img_i, path))

    # Create plot
    img = np.array(Image.open(path))
    plt.figure()
    fig, ax = plt.subplots(1)
    ax.imshow(img)

    # Draw bounding boxes and labels of detections
    if detections is not None:
        # Rescale boxes to original image
        detections = rescale_boxes(detections, 416, img.shape[:2])
        unique_labels = detections[:, -1].cpu().unique()
        n_cls_preds = len(unique_labels)
        bbox_colors = random.sample(colors, n_cls_preds)
        for x1, y1, x2, y2, conf, cls_conf, cls_pred in detections:

            print("\t+ Label: %s, Conf: %.5f" % (classes[int(cls_pred)], cls_conf.item()))

            box_w = x2 - x1
```

```
box_h = y2 - y1

color = bbox_colors[int(np.where(unique_labels == int(cls_pred))[0])]
# Create a Rectangle patch
bbox = patches.Rectangle((x1, y1), box_w, box_h, linewidth=2, edgecolor=color, facecolor='none')
# Add the bbox to the plot
ax.add_patch(bbox)
# Add label
plt.text(
    x1,
    y1,
    s=classes[int(cls_pred)],
    color="white",
    verticalalignment="top",
    bbox={"color": color, "pad": 0},
)

# Save generated image with detections
plt.axis("off")
plt.gca().xaxis.set_major_locator(NullLocator())
plt.gca().yaxis.set_major_locator(NullLocator())
filename = path.split("/)[-1].split(".")[0]
plt.savefig(f"output/{filename}.png", bbox_inches="tight", pad_inches=0.0)
plt.close()
```



```
+ Batch 0, Inference Time: 0:00:00.049708
+ Batch 1, Inference Time: 0:00:00.034727
+ Batch 2, Inference Time: 0:00:00.031957
+ Batch 3, Inference Time: 0:00:00.033261
+ Batch 4, Inference Time: 0:00:00.035032
+ Batch 5, Inference Time: 0:00:00.059134
+ Batch 6, Inference Time: 0:00:00.033763
+ Batch 7, Inference Time: 0:00:00.032069
+ Batch 8, Inference Time: 0:00:00.036263
```

Saving images:

```
(0) Image: 'data/samples/dog.jpg'
    + Label: dog, Conf: 0.99335
    + Label: bicycle, Conf: 0.99981
    + Label: truck, Conf: 0.94229
(1) Image: 'data/samples/eagle.jpg'
    + Label: bird, Conf: 0.99703
(2) Image: 'data/samples/field.jpg'
    + Label: person, Conf: 0.99996
    + Label: horse, Conf: 0.99977
    + Label: dog, Conf: 0.99409
(3) Image: 'data/samples/giraffe.jpg'
    + Label: giraffe, Conf: 0.99959
    + Label: zebra, Conf: 0.97958
(4) Image: 'data/samples/herd_of_horses.jpg'
    + Label: horse, Conf: 0.99459
    + Label: horse, Conf: 0.99352
    + Label: horse, Conf: 0.96845
    + Label: horse, Conf: 0.99478
(5) Image: 'data/samples/messi.jpg'
    + Label: person, Conf: 0.99993
    + Label: person, Conf: 0.99984
    + Label: person, Conf: 0.99996
(6) Image: 'data/samples/person.jpg'
    + Label: person, Conf: 0.99883
    + Label: dog, Conf: 0.99275
(7) Image: 'data/samples/room.jpg'
    + Label: chair, Conf: 0.99906
    + Label: chair, Conf: 0.96942
    + Label: clock, Conf: 0.99971
(8) Image: 'data/samples/street.jpg'
    + Label: car, Conf: 0.99977
    + Label: car, Conf: 0.99402
    + Label: car, Conf: 0.99841
    + Label: car, Conf: 0.99785
    + Label: car, Conf: 0.97907
    + Label: car, Conf: 0.95370
    + Label: traffic light, Conf: 0.99995
    + Label: car, Conf: 0.62254
<Figure size 432x288 with 0 Axes>
```

&lt;figure size 452x288 with 0 axes&gt;

```
from IPython.display import Image  
Image('/content/PyTorch-YOLOv3/data/samples/dog.jpg')
```



```
Image('/content/PyTorch-YOLOv3/output/dog.png')
```



