How to implement a YOLO (v3) object detector from scratch in PyTorch

<u>Part 1: Understanding How YOLO works</u> (<u>https://blog.paperspace.com/how-to-implement-a-yolo-object-detector-in-pytorch/</u>)

Github repo (https://github.com/ayooshkathuria/YOLO v3 tutorial from scratch)

<u>Part 2: Creating the layers of the network architecture</u> (https://blog.paperspace.com/how-to-implement-a-yolo-v3-object-detector-from-scratch-in-pytorch-part-2/)

```
In [1]: # mkdir cfg
# cd cfg
# wget https://raw.githubusercontent.com/pjreddie/darknet/master/cf
g/yolov3.cfg
```

```
In [2]: from __future__ import division

import time
import torch
import torch.nn as nn
import torch.nn.functional as F
from torch.autograd import Variable
import numpy as np
import cv2
import argparse
import os
import os.path as osp
import pickle as pkl
import pandas as pd
import random
```

```
In [3]: def parse cfg(cfgfile):
            Takes a configuration file
            Returns a list of blocks. Each blocks describes a block in the
        neural
            network to be built. Block is represented as a dictionary in th
        e list
            file = open(cfgfile, 'r')
            lines = file.read().split('\n')
                                                                   # store
        the lines in a list
            lines = [x \text{ for } x \text{ in lines if } len(x) > 0]
                                                                   # get re
        ad of the empty lines
            lines = [x \text{ for } x \text{ in lines if } x[0] != '#']
                                                                   # get ri
        d of comments
            lines = [x.rstrip().lstrip() for x in lines]
                                                                   # get ri
        d of fringe whitespaces
            block = \{\}
            blocks = []
            for line in lines:
                if line[0] == "[": # This marks the start of
        a new block
                    if len(block) != 0:
                                                 # If block is not empty, i
        mplies it is storing values of previous block.
                        blocks.append(block) # add it the blocks list
                        block = {}
                                                 # re-init the block
                    block["type"] = line[1:-1].rstrip()
                else:
                    key,value = line.split("=")
                    block[key.rstrip()] = value.lstrip()
            blocks.append(block)
            return blocks
In [4]: class EmptyLayer(nn.Module):
            def init (self):
                super(EmptyLayer, self). init ()
In [5]: class DetectionLayer(nn.Module):
            def init (self, anchors):
                super(DetectionLayer, self).__init__()
                self.anchors = anchors
In [6]: def create modules(blocks):
            net info = blocks[0] #Captures the information about the in
        put and pre-processing
            module list = nn.ModuleList()
```

```
prev filters = 3
    output filters = []
    for index, x in enumerate(blocks[1:]):
        module = nn.Sequential()
        #check the type of block
        #create a new module for the block
        #append to module list
        if (x["type"] == "convolutional"):
            #Get the info about the layer
            activation = x["activation"]
            try:
                batch_normalize = int(x["batch_normalize"])
                bias = False
            except:
                batch normalize = 0
                bias = True
            filters= int(x["filters"])
            padding = int(x["pad"])
            kernel size = int(x["size"])
            stride = int(x["stride"])
            if padding:
                pad = (kernel size - 1) // 2
            else:
                pad = 0
            #Add the convolutional layer
            conv = nn.Conv2d(prev_filters, filters, kernel_size, st
ride, pad, bias = bias)
            module.add_module("conv_{0}".format(index), conv)
            #Add the Batch Norm Layer
            if batch normalize:
                bn = nn.BatchNorm2d(filters)
                module.add_module("batch_norm_{0}".format(index), b
n)
            #Check the activation.
            #It is either Linear or a Leaky ReLU for YOLO
            if activation == "leaky":
                activn = nn.LeakyReLU(0.1, inplace = True)
                module.add module("leaky {0}".format(index), activn
)
        #If it's an upsampling layer
        #We use Bilinear2dUpsampling
        elif (x["type"] == "upsample"):
            stride = int(x["stride"])
            upsample = nn.Upsample(scale factor = 2, mode = "biline
```

```
ar")
            module.add module("upsample {}".format(index), upsample
)
        #If it is a route layer
        elif (x["type"] == "route"):
            x["layers"] = x["layers"].split(',')
            #Start of a route
            start = int(x["layers"][0])
            #end, if there exists one.
            try:
                end = int(x["layers"][1])
            except:
                end = 0
            #Positive anotation
            if start > 0:
                start = start - index
            if end > 0:
                end = end - index
            route = EmptyLayer()
            module.add_module("route_{0}".format(index), route)
            if end < 0:
                filters = output filters[index + start] + output fi
lters[index + end]
            else:
                filters= output_filters[index + start]
        #shortcut corresponds to skip connection
        elif x["type"] == "shortcut":
            shortcut = EmptyLayer()
            module.add module("shortcut {}".format(index), shortcut
        elif x["type"] == "yolo":
            mask = x["mask"].split(",")
            mask = [int(x) for x in mask]
            anchors = x["anchors"].split(",")
            anchors = [int(a) for a in anchors]
            anchors = [(anchors[i], anchors[i+1]) for i in range(0,
len(anchors),2)]
            anchors = [anchors[i] for i in mask]
            detection = DetectionLayer(anchors)
            module.add module("Detection {}".format(index), detecti
on)
        module_list.append(module)
        prev filters = filters
        output filters.append(filters)
    return (net info, module list)
```

```
In [7]: blocks = parse_cfg("cfg/yolov3.cfg")
    print(create_modules(blocks))
```

({'type': 'net', 'batch': '64', 'subdivisions': '16', 'width': '41

```
6', 'height': '416', 'channels': '3', 'momentum': '0.9', 'decay':
'0.0005', 'angle': '0', 'saturation': '1.5', 'exposure': '1.5', 'h
ue': '.1', 'learning rate': '0.001', 'burn in': '1000', 'max batch
es': '500200', 'policy': 'steps', 'steps': '400000,450000', 'scale
s': '.1,.1'}, ModuleList(
  (0): Sequential(
    (conv 0): Conv2d(3, 32, kernel size=(3, 3), stride=(1, 1), pad
ding=(1, 1), bias=False)
    (batch norm 0): BatchNorm2d(32, eps=1e-05, momentum=0.1, affin
e=True, track running stats=True)
    (leaky 0): LeakyReLU(negative slope=0.1, inplace=True)
  (1): Sequential(
    (conv 1): Conv2d(32, 64, kernel size=(3, 3), stride=(2, 2), pa
dding=(1, 1), bias=False)
    (batch norm 1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affin
e=True, track running stats=True)
    (leaky_1): LeakyReLU(negative_slope=0.1, inplace=True)
  )
  (2): Sequential(
    (conv 2): Conv2d(64, 32, kernel size=(1, 1), stride=(1, 1), bi
as=False)
    (batch norm 2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affin
e=True, track running stats=True)
    (leaky 2): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (3): Sequential(
    (conv 3): Conv2d(32, 64, kernel size=(3, 3), stride=(1, 1), pa
dding=(1, 1), bias=False)
    (batch norm 3): BatchNorm2d(64, eps=1e-05, momentum=0.1, affin
e=True, track running stats=True)
    (leaky 3): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (4): Sequential(
    (shortcut 4): EmptyLayer()
  (5): Sequential(
    (conv 5): Conv2d(64, 128, kernel size=(3, 3), stride=(2, 2), p
adding=(1, 1), bias=False)
    (batch_norm_5): BatchNorm2d(128, eps=1e-05, momentum=0.1, affi
ne=True, track_running_stats=True)
    (leaky 5): LeakyReLU(negative slope=0.1, inplace=True)
  (6): Sequential(
    (conv 6): Conv2d(128, 64, kernel size=(1, 1), stride=(1, 1), b
ias=False)
    (batch norm 6): BatchNorm2d(64, eps=1e-05, momentum=0.1, affin
e=True, track running stats=True)
    (leaky 6): LeakyReLU(negative slope=0.1, inplace=True)
  (7): Sequential(
    (conv_7): Conv2d(64, 128, kernel_size=(3, 3), stride=(1, 1), p
adding=(1, 1), bias=False)
```

```
(batch norm 7): BatchNorm2d(128, eps=1e-05, momentum=0.1, affi
ne=True, track running stats=True)
    (leaky 7): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (8): Sequential(
    (shortcut 8): EmptyLayer()
  (9): Sequential(
    (conv 9): Conv2d(128, 64, kernel size=(1, 1), stride=(1, 1), b
ias=False)
    (batch norm 9): BatchNorm2d(64, eps=1e-05, momentum=0.1, affin
e=True, track running stats=True)
    (leaky 9): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (10): Sequential(
    (conv 10): Conv2d(64, 128, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 10): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_10): LeakyReLU(negative_slope=0.1, inplace=True)
  )
  (11): Sequential(
    (shortcut 11): EmptyLayer()
  (12): Sequential(
    (conv 12): Conv2d(128, 256, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
    (batch norm 12): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 12): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (13): Sequential(
    (conv_13): Conv2d(256, 128, kernel_size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 13): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 13): LeakyReLU(negative slope=0.1, inplace=True)
  (14): Sequential(
    (conv 14): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 14): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 14): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (15): Sequential(
    (shortcut 15): EmptyLayer()
  (16): Sequential(
    (conv 16): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 16): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track_running_stats=True)
```

```
(leaky 16): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (17): Sequential(
    (conv 17): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 17): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 17): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (18): Sequential(
    (shortcut 18): EmptyLayer()
  (19): Sequential(
    (conv 19): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 19): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 19): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (20): Sequential(
    (conv 20): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 20): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 20): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (21): Sequential(
    (shortcut 21): EmptyLayer()
  (22): Sequential(
    (conv 22): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch_norm_22): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 22): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (23): Sequential(
    (conv 23): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch_norm_23): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track_running_stats=True)
    (leaky 23): LeakyReLU(negative slope=0.1, inplace=True)
  (24): Sequential(
    (shortcut 24): EmptyLayer()
  (25): Sequential(
    (conv 25): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 25): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 25): LeakyReLU(negative slope=0.1, inplace=True)
  )
```

```
(26): Sequential(
    (conv 26): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 26): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 26): LeakyReLU(negative slope=0.1, inplace=True)
  (27): Sequential(
    (shortcut 27): EmptyLayer()
  (28): Sequential(
    (conv 28): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 28): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 28): LeakyReLU(negative slope=0.1, inplace=True)
  (29): Sequential(
    (conv 29): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 29): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 29): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (30): Sequential(
    (shortcut 30): EmptyLayer()
  (31): Sequential(
    (conv 31): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 31): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_31): LeakyReLU(negative_slope=0.1, inplace=True)
  (32): Sequential(
    (conv 32): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 32): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_32): LeakyReLU(negative_slope=0.1, inplace=True)
  )
  (33): Sequential(
    (shortcut 33): EmptyLayer()
  (34): Sequential(
    (conv 34): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch_norm_34): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 34): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (35): Sequential(
    (conv_35): Conv2d(128, 256, kernel_size=(3, 3), stride=(1, 1),
```

```
padding=(1, 1), bias=False)
    (batch_norm_35): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 35): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (36): Sequential(
    (shortcut 36): EmptyLayer()
  )
  (37): Sequential(
    (conv 37): Conv2d(256, 512, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
    (batch norm 37): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 37): LeakyReLU(negative slope=0.1, inplace=True)
  (38): Sequential(
    (conv 38): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 38): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 38): LeakyReLU(negative slope=0.1, inplace=True)
  (39): Sequential(
    (conv 39): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 39): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track_running_stats=True)
    (leaky 39): LeakyReLU(negative slope=0.1, inplace=True)
  (40): Sequential(
    (shortcut 40): EmptyLayer()
  (41): Sequential(
    (conv 41): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 41): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 41): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (42): Sequential(
    (conv_42): Conv2d(256, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 42): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_42): LeakyReLU(negative_slope=0.1, inplace=True)
  (43): Sequential(
    (shortcut_43): EmptyLayer()
  (44): Sequential(
    (conv 44): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch_norm_44): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
```

```
ine=True, track running stats=True)
    (leaky 44): LeakyReLU(negative slope=0.1, inplace=True)
  (45): Sequential(
    (conv 45): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 45): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 45): LeakyReLU(negative slope=0.1, inplace=True)
  (46): Sequential(
    (shortcut 46): EmptyLayer()
  (47): Sequential(
    (conv 47): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 47): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 47): LeakyReLU(negative slope=0.1, inplace=True)
  (48): Sequential(
    (conv 48): Conv2d(256, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 48): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 48): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (49): Sequential(
    (shortcut 49): EmptyLayer()
  (50): Sequential(
    (conv 50): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 50): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 50): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (51): Sequential(
    (conv 51): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch_norm_51): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 51): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (52): Sequential(
    (shortcut 52): EmptyLayer()
  )
  (53): Sequential(
    (conv 53): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 53): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_53): LeakyReLU(negative_slope=0.1, inplace=True)
```

```
)
  (54): Sequential(
    (conv 54): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 54): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 54): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (55): Sequential(
    (shortcut 55): EmptyLayer()
  )
  (56): Sequential(
    (conv 56): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 56): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 56): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (57): Sequential(
    (conv_57): Conv2d(256, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch_norm_57): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_57): LeakyReLU(negative_slope=0.1, inplace=True)
  (58): Sequential(
    (shortcut_58): EmptyLayer()
  (59): Sequential(
    (conv 59): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 59): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track_running_stats=True)
    (leaky 59): LeakyReLU(negative slope=0.1, inplace=True)
  (60): Sequential(
    (conv 60): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 60): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_60): LeakyReLU(negative_slope=0.1, inplace=True)
  )
  (61): Sequential(
    (shortcut 61): EmptyLayer()
  (62): Sequential(
    (conv 62): Conv2d(512, 1024, kernel size=(3, 3), stride=(2, 2)
, padding=(1, 1), bias=False)
    (batch norm 62): BatchNorm2d(1024, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 62): LeakyReLU(negative slope=0.1, inplace=True)
  (63): Sequential(
```

```
(conv 63): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1)
, bias=False)
    (batch norm 63): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 63): LeakyReLU(negative slope=0.1, inplace=True)
  (64): Sequential(
    (conv 64): Conv2d(512, 1024, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch norm 64): BatchNorm2d(1024, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 64): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (65): Sequential(
    (shortcut 65): EmptyLayer()
  (66): Sequential(
    (conv 66): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1)
, bias=False)
    (batch_norm_66): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 66): LeakyReLU(negative slope=0.1, inplace=True)
  (67): Sequential(
    (conv 67): Conv2d(512, 1024, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch norm 67): BatchNorm2d(1024, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 67): LeakyReLU(negative slope=0.1, inplace=True)
  (68): Sequential(
    (shortcut 68): EmptyLayer()
  (69): Sequential(
    (conv 69): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1)
, bias=False)
    (batch norm 69): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 69): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (70): Sequential(
    (conv 70): Conv2d(512, 1024, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch norm 70): BatchNorm2d(1024, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 70): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (71): Sequential(
    (shortcut 71): EmptyLayer()
  (72): Sequential(
    (conv 72): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1)
, bias=False)
```

```
(batch norm 72): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 72): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (73): Sequential(
    (conv 73): Conv2d(512, 1024, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch_norm_73): BatchNorm2d(1024, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 73): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (74): Sequential(
    (shortcut 74): EmptyLayer()
  )
  (75): Sequential(
    (conv 75): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1)
, bias=False)
    (batch norm 75): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_75): LeakyReLU(negative_slope=0.1, inplace=True)
  )
  (76): Sequential(
    (conv 76): Conv2d(512, 1024, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch norm 76): BatchNorm2d(1024, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 76): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (77): Sequential(
    (conv 77): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1)
, bias=False)
    (batch norm 77): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track_running_stats=True)
    (leaky 77): LeakyReLU(negative slope=0.1, inplace=True)
  (78): Sequential(
    (conv 78): Conv2d(512, 1024, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch norm 78): BatchNorm2d(1024, eps=1e-05, momentum=0.1, af
fine=True, track_running_stats=True)
    (leaky_78): LeakyReLU(negative_slope=0.1, inplace=True)
  )
  (79): Sequential(
    (conv 79): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1)
, bias=False)
    (batch norm 79): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 79): LeakyReLU(negative slope=0.1, inplace=True)
  (80): Sequential(
    (conv 80): Conv2d(512, 1024, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch_norm_80): BatchNorm2d(1024, eps=1e-05, momentum=0.1, af
```

```
fine=True, track running stats=True)
    (leaky 80): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (81): Sequential(
    (conv 81): Conv2d(1024, 255, kernel size=(1, 1), stride=(1, 1)
)
  (82): Sequential(
    (Detection_82): DetectionLayer()
  (83): Sequential(
    (route 83): EmptyLayer()
  (84): Sequential(
    (conv 84): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 84): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 84): LeakyReLU(negative slope=0.1, inplace=True)
  (85): Sequential(
    (upsample 85): Upsample(scale_factor=2.0, mode=bilinear)
  (86): Sequential(
    (route 86): EmptyLayer()
  )
  (87): Sequential(
    (conv 87): Conv2d(768, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 87): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 87): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (88): Sequential(
    (conv 88): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 88): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 88): LeakyReLU(negative slope=0.1, inplace=True)
  (89): Sequential(
    (conv 89): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 89): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 89): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (90): Sequential(
    (conv 90): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 90): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_90): LeakyReLU(negative_slope=0.1, inplace=True)
```

```
)
  (91): Sequential(
    (conv 91): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 91): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 91): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (92): Sequential(
    (conv 92): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (batch norm 92): BatchNorm2d(512, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 92): LeakyReLU(negative slope=0.1, inplace=True)
  (93): Sequential(
    (conv 93): Conv2d(512, 255, kernel size=(1, 1), stride=(1, 1))
  )
  (94): Sequential(
    (Detection_94): DetectionLayer()
  )
  (95): Sequential(
    (route 95): EmptyLayer()
  (96): Sequential(
    (conv 96): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch norm 96): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky 96): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (97): Sequential(
    (upsample_97): Upsample(scale_factor=2.0, mode=bilinear)
  )
  (98): Sequential(
    (route 98): EmptyLayer()
  (99): Sequential(
    (conv 99): Conv2d(384, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
    (batch_norm_99): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
    (leaky_99): LeakyReLU(negative_slope=0.1, inplace=True)
  )
  (100): Sequential(
    (conv 100): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch_norm_100): BatchNorm2d(256, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 100): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (101): Sequential(
    (conv_101): Conv2d(256, 128, kernel_size=(1, 1), stride=(1, 1)
```

```
, bias=False)
    (batch norm 101): BatchNorm2d(128, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 101): LeakyReLU(negative slope=0.1, inplace=True)
  (102): Sequential(
    (conv 102): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch norm 102): BatchNorm2d(256, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 102): LeakyReLU(negative slope=0.1, inplace=True)
  (103): Sequential(
    (conv 103): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1)
, bias=False)
    (batch norm 103): BatchNorm2d(128, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 103): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (104): Sequential(
    (conv 104): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
    (batch norm 104): BatchNorm2d(256, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
    (leaky 104): LeakyReLU(negative slope=0.1, inplace=True)
  )
  (105): Sequential(
    (conv_105): Conv2d(256, 255, kernel_size=(1, 1), stride=(1, 1)
)
  (106): Sequential(
    (Detection 106): DetectionLayer()
  )
))
```

Part 3: Implementing the the forward pass of the network (https://blog.paperspace.com/how-to-implement-a-yolo-v3-object-detector-from-scratch-in-pytorch-part-3/)

```
In [8]: class Darknet(nn.Module):
    def __init__(self, cfgfile):
        super(Darknet, self).__init__()
        self.blocks = parse_cfg(cfgfile)
        self.net_info, self.module_list = create_modules(self.block
s)

def forward(self, x, CUDA):
    modules = self.blocks[1:]
    outputs = {} #We cache the outputs for the route layer
```

```
write = 0
        for i, module in enumerate(modules):
            module_type = (module["type"])
            if module type == "convolutional" or module type == "up
sample":
                x = self.module list[i](x)
            elif module type == "route":
                layers = module["layers"]
                layers = [int(a) for a in layers]
                if (layers[0]) > 0:
                    layers[0] = layers[0] - i
                if len(layers) == 1:
                    x = outputs[i + (layers[0])]
                else:
                    if (layers[1]) > 0:
                        layers[1] = layers[1] - i
                    map1 = outputs[i + layers[0]]
                    map2 = outputs[i + layers[1]]
                    x = torch.cat((map1, map2), 1)
            elif module type == "shortcut":
                from_ = int(module["from"])
                x = outputs[i-1] + outputs[i+from]
            elif module type == 'yolo':
                anchors = self.module list[i][0].anchors
                #Get the input dimensions
                inp dim = int (self.net info["height"])
                #Get the number of classes
                num classes = int (module["classes"])
                #Transform
                x = x.data
                x = predict transform(x, inp dim, anchors, num clas
ses, CUDA)
                if not write:
                                           #if no collector has bee
n intialised.
                    detections = x
                    write = 1
                else:
                    detections = torch.cat((detections, x), 1)
            outputs[i] = x
```

```
return detections
    def load weights(self, weightfile):
        #Open the weights file
        fp = open(weightfile, "rb")
        #The first 5 values are header information
        # 1. Major version number
        # 2. Minor Version Number
        # 3. Subversion number
        # 4,5. Images seen by the network (during training)
        header = np.fromfile(fp, dtype = np.int32, count = 5)
        self.header = torch.from numpy(header)
        self.seen = self.header[3]
        weights = np.fromfile(fp, dtype = np.float32)
        ptr = 0
        for i in range(len(self.module list)):
            module_type = self.blocks[i + 1]["type"]
            #If module type is convolutional load weights
            #Otherwise ignore.
            if module type == "convolutional":
                model = self.module list[i]
                try:
                    batch normalize = int(self.blocks[i+1]["batch n
ormalize"])
                except:
                    batch normalize = 0
                conv = model[0]
                if (batch normalize):
                    bn = model[1]
                    #Get the number of weights of Batch Norm Layer
                    num_bn_biases = bn.bias.numel()
                    #Load the weights
                    bn biases = torch.from numpy(weights[ptr:ptr +
num bn biases])
                    ptr += num bn biases
                    bn weights = torch.from numpy(weights[ptr: ptr
+ num bn biases])
                    ptr += num bn biases
                    bn running mean = torch.from numpy(weights[ptr:
ptr + num bn biases])
                    ptr += num_bn_biases
```

```
bn running var = torch.from numpy(weights[ptr:
ptr + num bn biases])
                    ptr += num bn biases
                    #Cast the loaded weights into dims of model wei
ghts.
                    bn biases = bn biases.view as(bn.bias.data)
                    bn weights = bn weights.view as(bn.weight.data)
                    bn running mean = bn running mean.view as(bn.ru
nning mean)
                    bn running var = bn running var.view as(bn.runn
ing var)
                    #Copy the data to model
                    bn.bias.data.copy (bn biases)
                    bn.weight.data.copy (bn weights)
                    bn.running mean.copy (bn running mean)
                    bn.running var.copy (bn running var)
                else:
                    #Number of biases
                    num biases = conv.bias.numel()
                    #Load the weights
                    conv biases = torch.from numpy(weights[ptr: ptr
+ num biases])
                    ptr = ptr + num biases
                    #reshape the loaded weights according to the di
ms of the model weights
                    conv biases = conv biases.view as(conv.bias.dat
a)
                    #Finally copy the data
                    conv.bias.data.copy (conv biases)
                #Let us load the weights for the Convolutional laye
rs
                num_weights = conv.weight.numel()
                #Do the same as above for weights
                conv weights = torch.from_numpy(weights[ptr:ptr+num
weights])
                ptr = ptr + num_weights
                conv weights = conv weights.view as(conv.weight.dat
a)
                conv.weight.data.copy (conv weights)
```

```
In [9]: def predict transform(prediction, inp dim, anchors, num classes, CU
        DA = True):
            batch size = prediction.size(0)
            stride = inp dim // prediction.size(2)
            grid size = inp dim // stride
            bbox attrs = 5 + num classes
            num anchors = len(anchors)
            prediction = prediction.view(batch size, bbox attrs*num anchors
        , grid_size*grid size)
            prediction = prediction.transpose(1,2).contiguous()
            prediction = prediction.view(batch size, grid size*grid size*nu
        m anchors, bbox attrs)
            anchors = [(a[0]/stride, a[1]/stride) for a in anchors]
            #Sigmoid the centre X, centre Y. and object confidence
            prediction[:,:,0] = torch.sigmoid(prediction[:,:,0])
            prediction[:,:,1] = torch.sigmoid(prediction[:,:,1])
            prediction[:,:,4] = torch.sigmoid(prediction[:,:,4])
            #Add the center offsets
            grid = np.arange(grid size)
            a,b = np.meshgrid(grid, grid)
            x 	ext{ offset} = torch.FloatTensor(a).view(-1,1)
            y offset = torch.FloatTensor(b).view(-1,1)
            if CUDA:
                x offset = x offset.cuda()
                y offset = y offset.cuda()
            x y offset = torch.cat((x offset, y offset), 1).repeat(1,num an
        chors).view(-1,2).unsqueeze(0)
            prediction[:,:,:2] += x_y_offset
            #log space transform height and the width
            anchors = torch.FloatTensor(anchors)
            if CUDA:
                anchors = anchors.cuda()
            anchors = anchors.repeat(grid size*grid size, 1).unsqueeze(0)
            prediction[:,:,2:4] = torch.exp(prediction[:,:,2:4])*anchors
            prediction[:,:,5: 5 + num classes] = torch.sigmoid((prediction[
        :,:, 5 : 5 + num classes]))
            prediction[:,:,:4] *= stride
            return prediction
```

```
In [11]: def get_test_input():
    img = cv2.imread("img/dog-cycle-car.png")
    img = cv2.resize(img, (416, 416))  #Resize to the input
    dimension
    img_ = img[:,:,::-1].transpose((2,0,1))  #BGR -> RGB | H X W C
    -> C X H X W
    img_ = img_[np.newaxis,:,:,:]/255.0  #Add a channel at 0 (
    for batch) | Normalise
    img_ = torch.from_numpy(img_).float()  #Convert to float
    img_ = Variable(img_)  #Convert to Variable
    return img_
```

```
In [12]: model = Darknet("cfg/yolov3.cfg")
    inp = get_test_input()
    print(inp.shape)
    pred = model(inp, torch.cuda.is_available())
    print (pred)
```

/Users/chanho/miniconda3/envs/yolo/lib/python3.8/site-packages/tor ch/nn/functional.py:2503: UserWarning: Default upsampling behavior when mode=bilinear is changed to align_corners=False since 0.4.0. Please specify align_corners=True if the old behavior is desired. See the documentation of nn.Upsample for details.

warnings.warn("Default upsampling behavior when mode={} is chang
ed "

```
In [13]: # wget https://pjreddie.com/media/files/yolov3.weights
In [14]: model = Darknet("cfg/yolov3.cfg")
    model.load weights("yolov3.weights")
```

```
In [15]: def bbox iou(box1, box2):
             Returns the IoU of two bounding boxes
             ......
             #Get the coordinates of bounding boxes
             b1 x1, b1 y1, b1 x2, b1 y2 = box1[:,0], box1[:,1], box1[:,2], b
         ox1[:,3]
             b2 x1, b2 y1, b2 x2, b2 y2 = box2[:,0], box2[:,1], box2[:,2], b
         ox2[:,3]
             #get the corrdinates of the intersection rectangle
             inter rect x1 = torch.max(b1 x1, b2 x1)
             inter rect y1 = torch.max(b1 y1, b2 y1)
             inter_rect_x2 = torch.min(b1_x2, b2 x2)
             inter_rect_y2 = torch.min(b1_y2, b2_y2)
             #Intersection area
             inter area = torch.clamp(inter rect x2 - inter rect x1 + 1, min
         =0) * torch.clamp(inter_rect_y2 - inter_rect_y1 + 1, min=0)
             #Union Area
             b1 area = (b1 x2 - b1 x1 + 1)*(b1 y2 - b1 y1 + 1)
             b2 area = (b2 x2 - b2 x1 + 1)*(b2 y2 - b2 y1 + 1)
             iou = inter area / (b1_area + b2_area - inter_area)
             return iou
```

<u>Part 4: Objectness score thresholding and Non-maximum suppression (https://blog.paperspace.com/how-to-implement-a-yolo-v3-object-detector-from-scratch-in-pytorch-part-4/)</u>

```
write = False
    for ind in range(batch size):
                                         #image Tensor
        image pred = prediction[ind]
       #confidence threshholding
       #NMS
        max conf, max conf score = torch.max(image pred[:,5:5+ num
classes], 1)
        max conf = max conf.float().unsqueeze(1)
        max conf score = max conf score.float().unsqueeze(1)
        seq = (image pred[:,:5], max conf, max conf score)
        image pred = torch.cat(seq, 1)
        non zero ind = (torch.nonzero(image pred[:,4]))
        try:
            image pred = image pred[non zero ind.squeeze(),:].view
(-1,7)
        except:
            continue
        if image_pred_.shape[0] == 0:
            continue
        #Get the various classes detected in the image
        img classes = unique(image pred [:,-1]) # -1 index holds t
he class index
        for cls in img classes:
            #perform NMS
            #get the detections with one particular class
            cls_mask = image_pred_*(image_pred_[:,-1] == cls).float
().unsqueeze(1)
            class mask ind = torch.nonzero(cls mask[:,-2]).squeeze(
)
            image pred class = image pred [class mask ind].view(-1,
7)
            #sort the detections such that the entry with the maxim
um objectness
            #confidence is at the top
            conf sort index = torch.sort(image pred class[:,4], des
cending = True )[1]
            image pred class = image pred class[conf sort index]
            idx = image pred class.size(0) #Number of detections
```

```
for i in range(idx):
                #Get the IOUs of all boxes that come after the one
we are looking at
                #in the loop
                try:
                    ious = bbox iou(image pred class[i].unsqueeze(0
), image pred class[i+1:])
                except ValueError:
                    break
                except IndexError:
                    break
                #Zero out all the detections that have IoU > treshh
old
                iou mask = (ious < nms conf).float().unsqueeze(1)</pre>
                image pred class[i+1:] *= iou mask
                #Remove the non-zero entries
                non zero ind = torch.nonzero(image pred class[:,4])
.squeeze()
                image pred class = image pred class[non zero ind].v
iew(-1,7)
            batch_ind = image_pred_class.new(image_pred_class.size(
0), 1).fill (ind)
                   #Repeat the batch id for as many detections
of the class cls in the image
            seq = batch ind, image pred class
            if not write:
                output = torch.cat(seq,1)
                write = True
                out = torch.cat(seq,1)
                output = torch.cat((output,out))
    try:
        return output
    except:
        return 0
```

<u>Part 5: Designing the input and the output pipelines</u> (<u>https://blog.paperspace.com/how-to-implement-a-yolo-v3-object-detector-from-scratch-in-pytorch-part-5/</u>)

```
In [17]: | img dir = 'img'
         det dir = 'det'
         batch size = 1
         confidence = 0.5
         nms thesh = 0.4
         cfgfile = 'cfg/yolov3.cfg'
         weightsfile = 'yolov3.weights'
         start = 0
         reso = 416
         CUDA = torch.cuda.is available()
In [18]: # mkdir data
         # cd data
         # wget https://raw.githubusercontent.com/ayooshkathuria/YOLO_v3_tut
         orial from scratch/master/data/coco.names
In [19]: | def load_classes(namesfile):
             fp = open(namesfile, "r")
             names = fp.read().split("\n")[:-1]
             return names
In [20]: | num_classes = 80
                              #For COCO
         classes = load classes("data/coco.names")
In [21]: | #Set up the neural network
         print("Loading network....")
         model = Darknet(cfgfile)
         model.load weights(weightsfile)
         print("Network successfully loaded")
         model.net info["height"] = reso
         inp dim = int(model.net info["height"])
         assert inp dim % 32 == 0
         assert inp dim > 32
         #If there's a GPU availible, put the model on GPU
         if CUDA:
             model.cuda()
         #Set the model in evaluation mode
         model.eval()
         Loading network.....
         Network successfully loaded
Out[21]: Darknet(
           (module list): ModuleList(
             (0): Sequential(
                (conv_0): Conv2d(3, 32, kernel_size=(3, 3), stride=(1, 1), p
         adding=(1, 1), bias=False)
               (batch norm 0): BatchNorm2d(32, eps=1e-05, momentum=0.1, aff
```

```
ine=True, track running stats=True)
      (leaky 0): LeakyReLU(negative slope=0.1, inplace=True)
    (1): Sequential(
      (conv 1): Conv2d(32, 64, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (batch norm 1): BatchNorm2d(64, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (leaky 1): LeakyReLU(negative slope=0.1, inplace=True)
    (2): Sequential(
      (conv 2): Conv2d(64, 32, kernel size=(1, 1), stride=(1, 1),
bias=False)
      (batch norm 2): BatchNorm2d(32, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (leaky 2): LeakyReLU(negative_slope=0.1, inplace=True)
    (3): Sequential(
      (conv 3): Conv2d(32, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (batch norm 3): BatchNorm2d(64, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (leaky 3): LeakyReLU(negative slope=0.1, inplace=True)
    (4): Sequential(
      (shortcut 4): EmptyLayer()
    (5): Sequential(
      (conv_5): Conv2d(64, 128, kernel_size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (batch norm 5): BatchNorm2d(128, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
      (leaky 5): LeakyReLU(negative slope=0.1, inplace=True)
    (6): Sequential(
      (conv 6): Conv2d(128, 64, kernel size=(1, 1), stride=(1, 1),
bias=False)
      (batch norm 6): BatchNorm2d(64, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (leaky 6): LeakyReLU(negative slope=0.1, inplace=True)
    (7): Sequential(
      (conv 7): Conv2d(64, 128, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (batch_norm_7): BatchNorm2d(128, eps=1e-05, momentum=0.1, af
fine=True, track running stats=True)
      (leaky 7): LeakyReLU(negative slope=0.1, inplace=True)
    (8): Sequential(
      (shortcut 8): EmptyLayer()
    (9): Sequential(
      (conv 9): Conv2d(128, 64, kernel size=(1, 1), stride=(1, 1),
```

```
bias=False)
      (batch norm 9): BatchNorm2d(64, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (leaky 9): LeakyReLU(negative slope=0.1, inplace=True)
    (10): Sequential(
      (conv 10): Conv2d(64, 128, kernel size=(3, 3), stride=(1, 1)
, padding=(1, 1), bias=False)
      (batch norm 10): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 10): LeakyReLU(negative slope=0.1, inplace=True)
    (11): Sequential(
      (shortcut 11): EmptyLayer()
    (12): Sequential(
      (conv 12): Conv2d(128, 256, kernel size=(3, 3), stride=(2, 2
), padding=(1, 1), bias=False)
      (batch norm 12): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 12): LeakyReLU(negative slope=0.1, inplace=True)
    (13): Sequential(
      (conv 13): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 13): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 13): LeakyReLU(negative slope=0.1, inplace=True)
    (14): Sequential(
      (conv 14): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 14): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 14): LeakyReLU(negative slope=0.1, inplace=True)
    (15): Sequential(
      (shortcut 15): EmptyLayer()
    (16): Sequential(
      (conv 16): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 16): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 16): LeakyReLU(negative slope=0.1, inplace=True)
    (17): Sequential(
      (conv 17): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 17): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 17): LeakyReLU(negative slope=0.1, inplace=True)
    )
```

```
(18): Sequential(
      (shortcut 18): EmptyLayer()
    (19): Sequential(
      (conv 19): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 19): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 19): LeakyReLU(negative slope=0.1, inplace=True)
    )
    (20): Sequential(
      (conv 20): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 20): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 20): LeakyReLU(negative slope=0.1, inplace=True)
    (21): Sequential(
      (shortcut 21): EmptyLayer()
    (22): Sequential(
      (conv 22): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 22): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 22): LeakyReLU(negative slope=0.1, inplace=True)
    (23): Sequential(
      (conv_23): Conv2d(128, 256, kernel_size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 23): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 23): LeakyReLU(negative slope=0.1, inplace=True)
    (24): Sequential(
      (shortcut 24): EmptyLayer()
    (25): Sequential(
      (conv_25): Conv2d(256, 128, kernel_size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 25): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 25): LeakyReLU(negative slope=0.1, inplace=True)
    (26): Sequential(
      (conv 26): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 26): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 26): LeakyReLU(negative slope=0.1, inplace=True)
    (27): Sequential(
      (shortcut_27): EmptyLayer()
```

```
)
    (28): Sequential(
      (conv 28): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 28): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 28): LeakyReLU(negative slope=0.1, inplace=True)
    )
    (29): Sequential(
      (conv 29): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 29): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 29): LeakyReLU(negative slope=0.1, inplace=True)
    (30): Sequential(
      (shortcut 30): EmptyLayer()
    (31): Sequential(
      (conv 31): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 31): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 31): LeakyReLU(negative slope=0.1, inplace=True)
    (32): Sequential(
      (conv 32): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 32): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 32): LeakyReLU(negative slope=0.1, inplace=True)
    (33): Sequential(
      (shortcut 33): EmptyLayer()
    (34): Sequential(
      (conv 34): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 34): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky_34): LeakyReLU(negative_slope=0.1, inplace=True)
    )
    (35): Sequential(
      (conv 35): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 35): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 35): LeakyReLU(negative slope=0.1, inplace=True)
    (36): Sequential(
      (shortcut 36): EmptyLayer()
    (37): Sequential(
```

```
(conv 37): Conv2d(256, 512, kernel size=(3, 3), stride=(2, 2
), padding=(1, 1), bias=False)
      (batch norm 37): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track_running_stats=True)
      (leaky 37): LeakyReLU(negative slope=0.1, inplace=True)
    (38): Sequential(
      (conv 38): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 38): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 38): LeakyReLU(negative slope=0.1, inplace=True)
    (39): Sequential(
      (conv 39): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch_norm_39): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 39): LeakyReLU(negative slope=0.1, inplace=True)
    (40): Sequential(
      (shortcut 40): EmptyLayer()
    (41): Sequential(
      (conv 41): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 41): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky_41): LeakyReLU(negative_slope=0.1, inplace=True)
    (42): Sequential(
      (conv 42): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 42): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky_42): LeakyReLU(negative_slope=0.1, inplace=True)
    (43): Sequential(
      (shortcut 43): EmptyLayer()
    (44): Sequential(
      (conv_44): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 44): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 44): LeakyReLU(negative slope=0.1, inplace=True)
    (45): Sequential(
      (conv 45): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 45): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 45): LeakyReLU(negative slope=0.1, inplace=True)
```

```
(46): Sequential(
      (shortcut 46): EmptyLayer()
    (47): Sequential(
      (conv 47): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch_norm_47): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 47): LeakyReLU(negative slope=0.1, inplace=True)
    (48): Sequential(
      (conv_48): Conv2d(256, 512, kernel_size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 48): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 48): LeakyReLU(negative slope=0.1, inplace=True)
    (49): Sequential(
      (shortcut_49): EmptyLayer()
    (50): Sequential(
      (conv 50): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 50): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 50): LeakyReLU(negative slope=0.1, inplace=True)
    (51): Sequential(
      (conv 51): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 51): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track_running_stats=True)
      (leaky 51): LeakyReLU(negative slope=0.1, inplace=True)
    (52): Sequential(
      (shortcut 52): EmptyLayer()
    (53): Sequential(
      (conv 53): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 53): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky_53): LeakyReLU(negative_slope=0.1, inplace=True)
    (54): Sequential(
      (conv 54): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 54): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 54): LeakyReLU(negative slope=0.1, inplace=True)
    (55): Sequential(
```

```
(shortcut 55): EmptyLayer()
    (56): Sequential(
      (conv 56): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 56): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 56): LeakyReLU(negative slope=0.1, inplace=True)
    (57): Sequential(
      (conv 57): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch_norm_57): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 57): LeakyReLU(negative slope=0.1, inplace=True)
    (58): Sequential(
      (shortcut 58): EmptyLayer()
    (59): Sequential(
      (conv 59): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 59): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 59): LeakyReLU(negative slope=0.1, inplace=True)
    )
    (60): Sequential(
      (conv 60): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 60): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 60): LeakyReLU(negative slope=0.1, inplace=True)
    (61): Sequential(
      (shortcut 61): EmptyLayer()
    (62): Sequential(
      (conv 62): Conv2d(512, 1024, kernel size=(3, 3), stride=(2,
2), padding=(1, 1), bias=False)
      (batch norm 62): BatchNorm2d(1024, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
      (leaky 62): LeakyReLU(negative slope=0.1, inplace=True)
    (63): Sequential(
      (conv 63): Conv2d(1024, 512, kernel size=(1, 1), stride=(1,
1), bias=False)
      (batch norm 63): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 63): LeakyReLU(negative slope=0.1, inplace=True)
    (64): Sequential(
      (conv_64): Conv2d(512, 1024, kernel_size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
```

```
(batch norm 64): BatchNorm2d(1024, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
      (leaky 64): LeakyReLU(negative slope=0.1, inplace=True)
    (65): Sequential(
      (shortcut 65): EmptyLayer()
    (66): Sequential(
      (conv 66): Conv2d(1024, 512, kernel size=(1, 1), stride=(1,
1), bias=False)
      (batch norm 66): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 66): LeakyReLU(negative slope=0.1, inplace=True)
    )
    (67): Sequential(
      (conv 67): Conv2d(512, 1024, kernel size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
      (batch norm 67): BatchNorm2d(1024, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
      (leaky_67): LeakyReLU(negative_slope=0.1, inplace=True)
    )
    (68): Sequential(
      (shortcut 68): EmptyLayer()
    (69): Sequential(
      (conv 69): Conv2d(1024, 512, kernel size=(1, 1), stride=(1,
1), bias=False)
      (batch norm 69): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 69): LeakyReLU(negative slope=0.1, inplace=True)
    (70): Sequential(
      (conv_70): Conv2d(512, 1024, kernel_size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
      (batch norm 70): BatchNorm2d(1024, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
      (leaky 70): LeakyReLU(negative slope=0.1, inplace=True)
    (71): Sequential(
      (shortcut_71): EmptyLayer()
    (72): Sequential(
      (conv 72): Conv2d(1024, 512, kernel size=(1, 1), stride=(1,
1), bias=False)
      (batch norm 72): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 72): LeakyReLU(negative slope=0.1, inplace=True)
    (73): Sequential(
      (conv 73): Conv2d(512, 1024, kernel size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
      (batch norm 73): BatchNorm2d(1024, eps=1e-05, momentum=0.1,
affine=True, track_running_stats=True)
```

```
(leaky 73): LeakyReLU(negative slope=0.1, inplace=True)
    (74): Sequential(
      (shortcut_74): EmptyLayer()
    (75): Sequential(
      (conv 75): Conv2d(1024, 512, kernel size=(1, 1), stride=(1,
1), bias=False)
      (batch norm 75): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 75): LeakyReLU(negative slope=0.1, inplace=True)
    (76): Sequential(
      (conv 76): Conv2d(512, 1024, kernel size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
      (batch norm 76): BatchNorm2d(1024, eps=1e-05, momentum=0.1,
affine=True, track_running_stats=True)
      (leaky_76): LeakyReLU(negative_slope=0.1, inplace=True)
    )
    (77): Sequential(
      (conv 77): Conv2d(1024, 512, kernel size=(1, 1), stride=(1,
1), bias=False)
      (batch norm 77): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 77): LeakyReLU(negative slope=0.1, inplace=True)
    )
    (78): Sequential(
      (conv 78): Conv2d(512, 1024, kernel size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
      (batch_norm_78): BatchNorm2d(1024, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
      (leaky 78): LeakyReLU(negative slope=0.1, inplace=True)
    (79): Sequential(
      (conv 79): Conv2d(1024, 512, kernel size=(1, 1), stride=(1,
1), bias=False)
      (batch norm 79): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 79): LeakyReLU(negative slope=0.1, inplace=True)
    (80): Sequential(
      (conv 80): Conv2d(512, 1024, kernel size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
      (batch norm 80): BatchNorm2d(1024, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
      (leaky 80): LeakyReLU(negative slope=0.1, inplace=True)
    )
    (81): Sequential(
      (conv 81): Conv2d(1024, 255, kernel size=(1, 1), stride=(1,
1))
    (82): Sequential(
      (Detection_82): DetectionLayer()
```

```
(83): Sequential(
      (route 83): EmptyLayer()
    (84): Sequential(
      (conv 84): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch_norm_84): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 84): LeakyReLU(negative slope=0.1, inplace=True)
    (85): Sequential(
      (upsample 85): Upsample(scale factor=2.0, mode=bilinear)
    (86): Sequential(
      (route 86): EmptyLayer()
    (87): Sequential(
      (conv 87): Conv2d(768, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch_norm_87): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky_87): LeakyReLU(negative_slope=0.1, inplace=True)
    (88): Sequential(
      (conv 88): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 88): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 88): LeakyReLU(negative slope=0.1, inplace=True)
    (89): Sequential(
      (conv_89): Conv2d(512, 256, kernel_size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 89): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 89): LeakyReLU(negative slope=0.1, inplace=True)
    (90): Sequential(
      (conv 90): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 90): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 90): LeakyReLU(negative slope=0.1, inplace=True)
    (91): Sequential(
      (conv 91): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 91): BatchNorm2d(256, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 91): LeakyReLU(negative slope=0.1, inplace=True)
    (92): Sequential(
```

```
(conv 92): Conv2d(256, 512, kernel size=(3, 3), stride=(1, 1
), padding=(1, 1), bias=False)
      (batch norm 92): BatchNorm2d(512, eps=1e-05, momentum=0.1, a
ffine=True, track_running_stats=True)
      (leaky 92): LeakyReLU(negative slope=0.1, inplace=True)
    (93): Sequential(
      (conv 93): Conv2d(512, 255, kernel size=(1, 1), stride=(1, 1
))
    (94): Sequential(
      (Detection 94): DetectionLayer()
    (95): Sequential(
      (route 95): EmptyLayer()
    (96): Sequential(
      (conv 96): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 96): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 96): LeakyReLU(negative slope=0.1, inplace=True)
    (97): Sequential(
      (upsample 97): Upsample(scale factor=2.0, mode=bilinear)
    (98): Sequential(
      (route 98): EmptyLayer()
    (99): Sequential(
      (conv 99): Conv2d(384, 128, kernel size=(1, 1), stride=(1, 1
), bias=False)
      (batch norm 99): BatchNorm2d(128, eps=1e-05, momentum=0.1, a
ffine=True, track running stats=True)
      (leaky 99): LeakyReLU(negative slope=0.1, inplace=True)
    )
    (100): Sequential(
      (conv 100): Conv2d(128, 256, kernel size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
      (batch norm 100): BatchNorm2d(256, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
      (leaky 100): LeakyReLU(negative slope=0.1, inplace=True)
    (101): Sequential(
      (conv 101): Conv2d(256, 128, kernel size=(1, 1), stride=(1,
1), bias=False)
      (batch norm 101): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
      (leaky 101): LeakyReLU(negative slope=0.1, inplace=True)
    (102): Sequential(
      (conv 102): Conv2d(128, 256, kernel size=(3, 3), stride=(1,
1), padding=(1, 1), bias=False)
```

(batch norm 102): BatchNorm2d(256, eps=1e-05, momentum=0.1,

```
affine=True, track running stats=True)
               (leaky 102): LeakyReLU(negative slope=0.1, inplace=True)
             (103): Sequential(
               (conv 103): Conv2d(256, 128, kernel size=(1, 1), stride=(1,
         1), bias=False)
               (batch norm 103): BatchNorm2d(128, eps=1e-05, momentum=0.1,
         affine=True, track running stats=True)
               (leaky 103): LeakyReLU(negative slope=0.1, inplace=True)
             (104): Sequential(
               (conv 104): Conv2d(128, 256, kernel size=(3, 3), stride=(1,
         1), padding=(1, 1), bias=False)
               (batch norm 104): BatchNorm2d(256, eps=1e-05, momentum=0.1,
         affine=True, track running stats=True)
               (leaky 104): LeakyReLU(negative slope=0.1, inplace=True)
             (105): Sequential(
               (conv 105): Conv2d(256, 255, kernel size=(1, 1), stride=(1,
         1))
             (106): Sequential(
               (Detection 106): DetectionLayer()
           )
         )
In [22]: read dir = time.time()
         #Detection phase
         try:
             imlist = [osp.join(osp.realpath('.'), img dir, img) for img in
         os.listdir(img dir) if img.find("png") > -1 ]
         except NotADirectoryError:
             imlist = []
             imlist.append(osp.join(osp.realpath('.'), img dir))
         except FileNotFoundError:
             print ("No file or directory with the name {}".format(img dir))
             exit()
In [23]: txt = '/Users/chanho/Documents/GitHub/keras-YOLOv3/img/frame0700.pn
```

True

print(txt.find("png") > -1)

```
In [24]: print(imlist)
```

['/Users/chanho/Documents/GitLab/niceface/models/img/frame0900.png', '/Users/chanho/Documents/GitLab/niceface/models/img/zebra.png', '/Users/chanho/Documents/GitLab/niceface/models/img/frame1700.png', '/Users/chanho/Documents/GitLab/niceface/models/img/frame1900.png', '/Users/chanho/Documents/GitLab/niceface/models/img/frame1309.png', '/Users/chanho/Documents/GitLab/niceface/models/img/frame0600.png', '/Users/chanho/Documents/GitLab/niceface/models/img/frame1291.png', '/Users/chanho/Documents/GitLab/niceface/models/img/frame1441.png', '/Users/chanho/Documents/GitLab/niceface/models/img/dog-cycle-car.png', '/Users/chanho/Documents/GitLab/niceface/models/img/frame0700.png']

```
In [29]: #PyTorch Variables for images
    im_batches = list(map(prep_image, loaded_ims, [inp_dim for x in ran
        ge(len(imlist))]))

#List containing dimensions of original images
    im_dim_list = [(x.shape[1], x.shape[0]) for x in loaded_ims]
    im_dim_list = torch.FloatTensor(im_dim_list).repeat(1,2)

if CUDA:
    im_dim_list = im_dim_list.cuda()
In [301: leftover = 0
```

```
In [31]: def unique(tensor):
    tensor_np = tensor.cpu().numpy()
    unique_np = np.unique(tensor_np)
    unique_tensor = torch.from_numpy(unique_np)

    tensor_res = tensor.new(unique_tensor.shape)
    tensor_res.copy_(unique_tensor)
    return tensor_res
```

```
In [32]: write = 0
         start det loop = time.time()
         for i, batch in enumerate(im batches):
             #load the image
             start = time.time()
             if CUDA:
                 batch = batch.cuda()
             prediction = model(Variable(batch, volatile = True), CUDA)
             prediction = write results(prediction, confidence, num classes,
         nms conf = nms thesh)
             end = time.time()
             if type(prediction) == int:
                 for im num, image in enumerate(imlist[i*batch size: min((i
         + 1)*batch size, len(imlist))):
                     im id = i*batch size + im num
                    print("{0:20s} predicted in {1:6.3f} seconds".format(im
         age.split("/")[-1], (end - start)/batch size))
                    print("{0:20s} {1:s}".format("Objects Detected:", ""))
                 continue
             prediction[:,0] += i*batch size #transform the atribute from
         index in batch to index in imlist
             if not write:
                                              #If we have't initialised ou
         tput
                output = prediction
                write = 1
             else:
                 output = torch.cat((output,prediction))
             for im num, image in enumerate(imlist[i*batch size: min((i + 1
         )*batch size, len(imlist))]):
                 im id = i*batch size + im num
                 objs = [classes[int(x[-1])] for x in output if int(x[0]) ==
         im id]
                 print("{0:20s} predicted in {1:6.3f} seconds".format(image.
         split("/")[-1], (end - start)/batch size))
                print("{0:20s} {1:s}".format("Objects Detected:", " ".join(
         objs)))
                print("-----
             if CUDA:
                 torch.cuda.synchronize()
```

<ipython-input-32-e5724478fd54>:9: UserWarning: volatile was remov
ed and now has no effect. Use `with torch.no_grad():` instead.
 prediction = model(Variable(batch, volatile = True), CUDA)

predicted in 0.863 seconds frame0900.png Objects Detected: person person person person person per son person _____ zebra.png predicted in 0.801 seconds Objects Detected: zebra zebra zebra _____ frame1700.png predicted in 0.782 seconds
Objects Detected: person person person person
-----frame1900.png predicted in 0.774 seconds
Objects Detected: person person person person person person person per son person person frame1309.png predicted in 0.771 seconds
Objects Detected: person person person person person _____ frame0600.png predicted in 0.795 seconds
Objects Detected: person person person person person person per son train backpack _____ frame1291.png predicted in 0.808 seconds
Objects Detected: person person _____ frame1441.png predicted in 0.813 seconds
Objects Detected: person person person person dog-cycle-car.png predicted in 0.830 seconds
Objects Detected: bicycle truck dog _____ frame0700.png predicted in 0.780 seconds Objects Detected: person person person person _____

```
In [33]: try:
    output
    except NameError:
        print ("No detections were made")
        exit()
```

```
In [34]: im dim list = torch.index select(im dim list, 0, output[:,0].long()
         scaling factor = torch.min(inp dim/im dim list,1)[0].view(-1,1)
         output[:,[1,3]] -= (inp_dim - scaling_factor*im dim list[:,0].view(
         -1,1))/2
         output[:,[2,4]] -= (inp_dim - scaling_factor*im_dim_list[:,1].view(
         -1,1))/2
         output[:,1:5] /= scaling factor
In [35]: | for i in range(output.shape[0]):
             output[i, [1,3]] = torch.clamp(output[i, [1,3]], 0.0, im dim li
             output[i, [2,4]] = torch.clamp(output[i, [2,4]], 0.0, im dim li
         st[i,1])
In [36]: | class load = time.time()
         colors = pkl.load(open("data/pallete", "rb"))
In [37]: draw = time.time()
         def write(x, results, color):
             c1 = tuple(x[1:3].int())
             c2 = tuple(x[3:5].int())
             img = results[int(x[0])]
             cls = int(x[-1])
             label = "{0}".format(classes[cls])
             cv2.rectangle(img, c1, c2,color, 1)
             t size = cv2.getTextSize(label, cv2.FONT HERSHEY PLAIN, 1 , 1)[
         0]
             c2 = c1[0] + t_size[0] + 3, c1[1] + t_size[1] + 4
             cv2.rectangle(img, c1, c2,color, -1)
             cv2.putText(img, label, (c1[0], c1[1] + t size[1] + 4), cv2.FON
         T_HERSHEY_PLAIN, 1, [225,255,255], 1);
             return img
In [38]: list(map(lambda x: write(x, loaded_ims, colors[0]), output))
         det names = pd.Series(imlist).apply(lambda x: "{}/det {}".format(de
         t dir,x.split("/")[-1]))
         list(map(cv2.imwrite, det names, loaded ims))
         end = time.time()
```

Here is a sample image of results.

